

User's Manual

Version 7.0

Ray D. Zimmerman Carlos E. Murillo-Sánchez

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Contents

1	Intr	roduction													
	1.1	1 Background													
	1.2	License and Terms of Use													
	1.3	Citing Matpower													
	1.4	MATPOWER Development													
	1.5	Sponsoring the Matpower Project													
2	Get	ting Started													
	2.1	System Requirements													
	2.2	Getting Matpower													
		2.2.1 Versioned Releases													
		2.2.2 Current Development Version													
		2.2.3 Matpower Docker Image													
	2.3	Installation													
	2.4	Running a Simulation													
		2.4.1 Preparing Case Input Data													
		2.4.2 Solving the Case													
		2.4.3 Accessing the Results													
		2.4.4 Setting Options													
	2.5	Documentation													
3	Mo	Modeling													
	3.1	Data Formats													
	3.2	Branches													
	3.3	Generators													
	3.4	Loads													
	3.5	Shunt Elements													
	3.6	Network Equations													
	3.7	DC Modeling													
4	Pov	Power Flow 32													
	4.1	AC Power Flow													
		4.1.1 Cartesian vs. Polar Coordinates for Voltage													
		4.1.2 Current vs. Power for Nodal Balance Equations													
	4.2	DC Power Flow													
	4.3	Distribution Power Flow													
	_,,	4.3.1 Radial Power Flow													

		4.3.2	Current Summation Method	38
		4.3.3	Power Summation Method	38
		4.3.4	Admittance Summation Method	39
		4.3.5	Handling PV Buses	40
	4.4	runpf	4	42
	4.5	Linear	Shift Factors	43
5	Cor	ntinuat	ion Power Flow 4	17
	5.1	Param	reterization	48
	5.2	Predic	t <mark>or</mark>	48
	5.3	Correc	etor	49
	5.4	Step I	ength Control	49
	5.5	Event	Detection and Location	50
	5.6	runcp	f	51
		5.6.1	CPF Callback Functions	54
		5.6.2	CPF Example	57
6	Opt	imal F	Power Flow 6	60
	6.1			60
		6.1.1		32
		6.1.2	$oldsymbol{arphi}$	32
	6.2	Standa		33
	6.3			64
		6.3.1	User-defined Variables	35
		6.3.2	User-defined Constraints	35
		6.3.3		66
	6.4	Standa	ard Extensions	38
		6.4.1	Piecewise Linear Costs	38
		6.4.2	Dispatchable Loads	70
		6.4.3		72
		6.4.4		72
	6.5	Solver		73
	6.6			74
7	Ext	ending	the OPF 8	30
	7.1	_		30
		7.1.1		30
		7.1.2		30
		7.1.3		32

		7.1.4	Additional Comments	33
	7.2	Callba	ack Functions	33
		7.2.1	User-defined Variables	34
		7.2.2	User-defined Costs	34
		7.2.3	User-defined Constraints	35
	7.3	Callba	ack Stages and Example	36
		7.3.1	ext2int Callback	37
		7.3.2	formulation Callback	38
		7.3.3	int2ext Callback	92
		7.3.4	printpf Callback	95
		7.3.5	savecase Callback	97
	7.4	Regist	tering the Callbacks	99
	7.5	Summ	nary)1
	7.6	Exam	ple Extensions)1
		7.6.1	Fixed Zonal Reserves)1
		7.6.2	Interface Flow Limits)3
		7.6.3	DC Transmission Lines)4
		7.6.4	OPF Soft Limits)7
8	Uni	t De-c	commitment Algorithm 11	4
9	Mis	cellan	eous Matpower Functions 11	.6
9	Mis 9.1			_
9			Output Functions	16
9		Input	/Output Functions	16 16
9		Input, 9.1.1	Output Functions 11 loadcase 11 savecase 11	l6 l6
9		Input, 9.1.1 9.1.2	/Output Functions 11 loadcase 11 savecase 11 cdf2mpc 11	16 16 16
9		Input, 9.1.1 9.1.2 9.1.3	/Output Functions 11 loadcase 11 savecase 11 cdf2mpc 11	16 16 16 17
9		Input, 9.1.1 9.1.2 9.1.3 9.1.4 9.1.5	/Output Functions 11 loadcase 11 savecase 11 cdf2mpc 11 psse2mpc 11	16 16 17 17
9	9.1	Input, 9.1.1 9.1.2 9.1.3 9.1.4 9.1.5	/Output Functions 11 loadcase 11 savecase 11 cdf2mpc 11 psse2mpc 11 save2psse 11	16 16 17 18
9	9.1	Input, 9.1.1 9.1.2 9.1.3 9.1.4 9.1.5 System	/Output Functions 11 loadcase 11 savecase 11 cdf2mpc 11 psse2mpc 11 save2psse 11 m Information 11	16 16 17 17 18
9	9.1	Input, 9.1.1 9.1.2 9.1.3 9.1.4 9.1.5 System 9.2.1 9.2.2	/Output Functions 11 loadcase 11 savecase 11 cdf2mpc 11 psse2mpc 11 save2psse 11 m Information 11 case_info 11	16 16 17 18 18
9	9.1	Input, 9.1.1 9.1.2 9.1.3 9.1.4 9.1.5 System 9.2.1 9.2.2	/Output Functions 11 loadcase 11 savecase 11 cdf2mpc 11 psse2mpc 11 save2psse 11 n Information 11 case_info 11 compare_case 11	16 16 17 18 18
9	9.1	Input, 9.1.1 9.1.2 9.1.3 9.1.4 9.1.5 System 9.2.1 9.2.2 9.2.3	/Output Functions 11 loadcase 11 savecase 11 cdf2mpc 11 psse2mpc 11 save2psse 11 m Information 11 case_info 11 compare_case 11 find_islands 11	16 16 17 18 18 18
9	9.1	Input, 9.1.1 9.1.2 9.1.3 9.1.4 9.1.5 System 9.2.1 9.2.2 9.2.3 9.2.4	/Output Functions 11 loadcase 11 savecase 11 cdf2mpc 11 psse2mpc 11 save2psse 11 n Information 11 case_info 11 compare_case 11 find_islands 11 get_losses 11	16 16 17 17 18 18 18
9	9.1	Input, 9.1.1 9.1.2 9.1.3 9.1.4 9.1.5 System 9.2.1 9.2.2 9.2.3 9.2.4 9.2.5	/Output Functions 11 loadcase 11 savecase 11 cdf2mpc 11 psse2mpc 11 save2psse 11 m Information 11 case_info 11 compare_case 11 find_islands 11 get_losses 11 margcost 12	16 16 17 18 18 18 19 19
9	9.1	Input, 9.1.1 9.1.2 9.1.3 9.1.4 9.1.5 System 9.2.1 9.2.2 9.2.3 9.2.4 9.2.5 9.2.6	/Output Functions 11 loadcase 11 savecase 11 cdf2mpc 11 psse2mpc 11 save2psse 11 m Information 11 case_info 11 compare_case 11 find_islands 11 get_losses 11 margcost 12 isload 12	16 16 17 17 18 18 18 19 20 20

	9.2.10	totcost
9.3	Modif	ying a Case
	9.3.1	extract_islands
	9.3.2	load2disp
	9.3.3	modcost
	9.3.4	scale_load
	9.3.5	apply_changes
	9.3.6	savechgtab
9.4	Conve	rsion between External and Internal Numbering
	9.4.1	ext2int, int2ext
	9.4.2	e2i_data, i2e_data
	9.4.3	e2i_field, i2e_field
9.5	Formi	ng Standard Power Systems Matrices
	9.5.1	makeB
	9.5.2	makeBdc
	9.5.3	makeJac
	9.5.4	makeLODF
	9.5.5	makePTDF
	9.5.6	makeYbus
9.6	Miscel	laneous
	9.6.1	define_constants
	9.6.2	feval_w_path
	9.6.3	have_fcn
	9.6.4	mpopt2qpopt
	9.6.5	mpver
	9.6.6	nested_struct_copy
10 Ack	nowle	dgments 134
Appen	dix A	MIPS – MATPOWER Interior Point Solver 135
		ple 1
	_	p <mark>le 2</mark>
A.3		ratic Programming Solver
A.4	•	l-Dual Interior Point Algorithm
	A.4.1	
	A.4.2	Problem Formulation and Lagrangian
	A.4.3	First Order Optimality Conditions
	A.4.4	Newton Step

Appendix B Data File Format				148
Appendix C Matpower Options C.1 Mapping of Old-Style Options to New-Style Options				1 54
C.1 Mapping of Old-Style Options to New-Style Options	•		•	170
Appendix D Matpower Files and Functions				174
D.1 Directory Layout and Documentation Files				174
D.2 Matpower Functions				176
D.3 Example Matpower Cases				188
D.4 Automated Test Suite				191
Appendix E MATPOWER Extras				195
Appendix F "Smart Market" Code				197
F.1 Handling Supply Shortfall				199
F.2 Example				199
F.3 Smartmarket Files and Functions				204
Appendix G Optional Packages				205
G.1 BPMPD_MEX – MEX interface for BPMPD				205
G.2 CLP – COIN-OR Linear Programming				205
G.3 CPLEX – High-performance LP and QP Solvers				206
G.4 GLPK – GNU Linear Programming Kit				207
G.5 Gurobi – High-performance LP and QP Solvers				207
G.6 IPOPT – Interior Point Optimizer				208
G.7 Artelys Knitro – Non-Linear Programming Solver				209
G.8 MINOPF – AC OPF Solver Based on MINOS				210
G.9 MOSEK – High-performance LP and QP Solvers				210
G.10 Optimization Toolbox – LP, QP, NLP and MILP Solvers				211
G.11 PARDISO – Parallel Sparse Direct and Multi-Recursive Iterativ	e I	inے		
ear Solvers				211
G.12 SDP_PF – Applications of a Semidefinite Programming Relaxat	ior	1 0	f	
the Power Flow Equations				212
G.13 TSPOPF – Three AC OPF Solvers by H. Wang				
Appendix H Release History				214
H.1 Pre 1.0 – released Jun 25, 1997				214
H.2 Version 1.0 – released Sep 17, 1997				
H.3 Version 1.0.1 – released Sep 19, 1997				

H.4	Version	2.0	– release	ed Dec	24,	1997		 						215
H.5	Version	3.0	– release	ed Feb	14,	2005		 						216
H.6	Version	3.2	– release	ed Sep	21,	2007		 						217
H.7	Version	4.0	– release	ed Feb	7, 2	011 .		 						219
H.8	Version	4.1	– release	ed Dec	: 14,	2011		 						222
H.9	Version	5.0	– release	ed Dec	e 17,	2014		 						223
H.10) Version	5.1	– release	ed Ma	r 20,	2015		 						228
H.11	Version	6.0	– release	ed Dec	: 16,	2016		 						231
H.12	2 Version	7.0	– release	ed Jun	20,	2019								236
Refere	nces													245

List of Figures

3-1	Branch Model	26
4-1	Oriented Ordering	37
4-2	Branch Representation: branch k between buses i (sending) and k	
	(receiving) and load demand and shunt admittances at both buses	37
5-1	Nose Curve of Voltage Magnitude at Bus 9	57
6-1	Relationship of w_i to r_i for $d_i = 1$ (linear option)	67
6-2	Relationship of w_i to r_i for $d_i = 2$ (quadratic option)	68
6-3	Constrained Cost Variable	69
6-4	Marginal Benefit or Bid Function	70
6-5	Total Cost Function for Negative Injection	71
6-6	Generator P - Q Capability Curve	73
7-1	Adding Constraints Across Subsets of Variables	91
7-2	DC Line Model	105
7-3	Equivalent "Dummy" Generators	105
7-4	Feasible Region for Branch Flow Violation Constraints	108
T ict	of Tables	
List	of Tables	
4-1	Power Flow Results	43
	Power Flow Results	43 45
4-1 4-2 4-3	Power Flow Options	43 45 46
4-2	Power Flow Options	45 46
4-2 4-3	Power Flow Options	45
4-2 4-3 5-1	Power Flow Options	45 46 52
4-2 4-3 5-1 5-2	Power Flow Options	45 46 52 53
4-2 4-3 5-1 5-2 5-3	Power Flow Options Power Flow Output Options Continuation Power Flow Results Continuation Power Flow Options Continuation Power Flow Callback Input Arguments	45 46 52 53 55
4-2 4-3 5-1 5-2 5-3 5-4	Power Flow Options Power Flow Output Options Continuation Power Flow Results Continuation Power Flow Options Continuation Power Flow Callback Input Arguments Continuation Power Flow Callback Output Arguments	45 46 52 53 55 56
4-2 4-3 5-1 5-2 5-3 5-4 5-5	Power Flow Options Power Flow Output Options Continuation Power Flow Results Continuation Power Flow Options Continuation Power Flow Callback Input Arguments Continuation Power Flow Callback Output Arguments Continuation Power Flow State	45 46 52 53 55 56 56
4-2 4-3 5-1 5-2 5-3 5-4 5-5 6-1	Power Flow Options Power Flow Output Options Continuation Power Flow Results Continuation Power Flow Options Continuation Power Flow Callback Input Arguments Continuation Power Flow Callback Output Arguments Continuation Power Flow State Optimal Power Flow Results	45 46 52 53 55 56 56 75
4-2 4-3 5-1 5-2 5-3 5-4 5-5 6-1 6-2	Power Flow Options Power Flow Output Options Continuation Power Flow Results Continuation Power Flow Options Continuation Power Flow Callback Input Arguments Continuation Power Flow Callback Output Arguments Continuation Power Flow State Optimal Power Flow Results Optimal Power Flow Solver Options Other OPF Options	45 46 52 53 55 56 56 75
4-2 4-3 5-1 5-2 5-3 5-4 5-5 6-1 6-2 6-3	Power Flow Options Power Flow Output Options Continuation Power Flow Results Continuation Power Flow Options Continuation Power Flow Callback Input Arguments Continuation Power Flow Callback Output Arguments Continuation Power Flow State Optimal Power Flow Results Optimal Power Flow Solver Options Other OPF Options OPF Output Options	45 46 52 53 55 56 56 75 77
4-2 4-3 5-1 5-2 5-3 5-4 5-5 6-1 6-2 6-3 6-4	Power Flow Options Power Flow Output Options Continuation Power Flow Results Continuation Power Flow Options Continuation Power Flow Callback Input Arguments Continuation Power Flow Callback Output Arguments Continuation Power Flow State Optimal Power Flow Results Optimal Power Flow Solver Options Other OPF Options	45 46 52 53 55 56 75 77 78 79
4-2 4-3 5-1 5-2 5-3 5-4 5-5 6-1 6-2 6-3 6-4 7-1	Power Flow Options Power Flow Output Options Continuation Power Flow Results Continuation Power Flow Options Continuation Power Flow Callback Input Arguments Continuation Power Flow Callback Output Arguments Continuation Power Flow State Optimal Power Flow Results Optimal Power Flow Solver Options Other OPF Options OPF Output Options User-defined Nonlinear Constraint Specification	45 46 52 53 55 56 56 75 77 78 79 82 87
4-2 4-3 5-1 5-2 5-3 5-4 5-5 6-1 6-2 6-3 6-4 7-1	Power Flow Options Power Flow Output Options Continuation Power Flow Results Continuation Power Flow Options Continuation Power Flow Callback Input Arguments Continuation Power Flow Callback Output Arguments Continuation Power Flow State Optimal Power Flow Results Optimal Power Flow Solver Options Other OPF Options OPF Output Options User-defined Nonlinear Constraint Specification Names Used by Implementation of OPF with Reserves	45 46 52 53 55 56 56 77 78 79 82
4-2 4-3 5-1 5-2 5-3 5-4 5-5 6-1 6-2 6-3 6-4 7-1 7-2 7-3	Power Flow Options Power Flow Output Options Continuation Power Flow Results Continuation Power Flow Options Continuation Power Flow Callback Input Arguments Continuation Power Flow Callback Output Arguments Continuation Power Flow State Optimal Power Flow Results Optimal Power Flow Results Optimal Power Flow Solver Options Other OPF Options OPF Output Options User-defined Nonlinear Constraint Specification Names Used by Implementation of OPF with Reserves Results for User-Defined Variables, Constraints and Costs	45 46 52 53 55 56 56 77 78 79 82 87 93
4-2 4-3 5-1 5-2 5-3 5-4 5-5 6-1 6-2 6-3 6-4 7-1 7-2 7-3 7-4	Power Flow Options Power Flow Output Options Continuation Power Flow Results Continuation Power Flow Options Continuation Power Flow Callback Input Arguments Continuation Power Flow Callback Output Arguments Continuation Power Flow State Optimal Power Flow Results Optimal Power Flow Results Optimal Power Flow Solver Options Other OPF Options OPF Output Options User-defined Nonlinear Constraint Specification Names Used by Implementation of OPF with Reserves Results for User-Defined Variables, Constraints and Costs Callback Functions	45 46 52 53 55 56 56 75 77 78 79 82 87 93 101

7-7	Input Data Structures for Interface Flow Limits
7-8	Output Data Structures for Interface Flow Limits
7-9	Soft Limit Formulation
7-10	Input Data Structures for OPF Soft Limits
7-11	Default Soft Limit Values
7-12	Possible Hard-Limit Modifications
7-13	Output Data Structures for OPF Soft Limits
9-1	Columns of chgtab
9-2	Values for CT_TABLE Column
9-3	Values for CT_CHGTYPE Column
9-4	Values for CT_COL Column
A-1	Input Arguments for mips
A-2	Output Arguments for mips
A-3	Options for mips
B-1	Bus Data (mpc.bus)
B-2	Generator Data (mpc.gen)
B-3	Branch Data (mpc.branch)
B-4	Generator Cost Data (mpc.gencost)
B-5	DC Line Data (mpc.dcline)
C-1	Top-Level Options
C-2	Power Flow Options
C-3	Continuation Power Flow Options
C-4	OPF Solver Options
C-5	General OPF Options
C-6	Power Flow and OPF Output Options
C-7	OPF Options for MIPS
C-8	OPF Options for CLP
	OPF Options for CPLEX
	OPF Options for fmincon
	OPF Options for GLPK
	OPF Options for Gurobi
	OPF Options for IPOPT
	OPF Options for Artelys Knitro
	OPF Options for MINOPF
	OPF Options for MOSEK
C-17	OPF Options for PDIPM
	OPF Options for TRALM
C-19	Old-Style to New-Style Option Mapping

D-1	Matpower Directory Layout and Documentation Files 175
D-2	Top-Level Simulation Functions
D-3	Input/Output Functions
D-4	Data Conversion Functions
D-5	Power Flow Functions
D-6	Continuation Power Flow Functions
D-7	OPF and Wrapper Functions
D-8	OPF Model Objects
D-9	Deprecated Copt_model Methods
	OPF Solver Functions
	Other OPF Functions
D-12	OPF User Callback Functions
D-13	Power Flow Derivative Functions
D-14	NLP, LP & QP Solver Functions
	Matrix Building Functions
D-16	Utility Functions
D-17	Other Functions
D-18	Small Transmission System Test Cases
D-19	Small Radial Distribution System Test Cases
D-20	ACTIV Synthetic Grid Test Cases
D-21	Polish System Test Cases
	PEGASE European System Test Cases
D-23	RTE French System Test Cases
	Automated Test Functions from MP-Test
D-25	MIPS Tests
D-26	Test Data
D-27	Miscellaneous Matpower Tests
	MATPOWER Power Flow, CPF and OPF Tests
F-1	Auction Types
F-2	Generator Offers
F-3	Load Bids
F-4	Generator Sales
F-5	Load Purchases
F_6	Smartmarket Files and Functions 204



1 Introduction

1.1 Background

MATPOWER [1,2] is a package of MATLAB® M-files for solving power flow and optimal power flow problems. It is intended as a simulation tool for researchers and educators that is easy to use and modify. MATPOWER is designed to give the best performance possible while keeping the code simple to understand and modify. The MATPOWER website can be found at:

https://matpower.org

MATPOWER was initially developed by Ray D. Zimmerman, Carlos E. Murillo-Sánchez and Deqiang Gan of PSERC¹ at Cornell University under the direction of Robert J. Thomas. The initial need for MATLAB-based power flow and optimal power flow code was born out of the computational requirements of the PowerWeb project². Many others have contributed to MATPOWER over the years and it continues to be developed and maintained under the direction of Ray Zimmerman.

Beginning with version 6, MATPOWER includes a framework for solving generalized steady-state electric power scheduling problems. This framework is known as MOST, for MATPOWER Optimal Scheduling Tool [3,4].

MOST can be used to solve problems as simple as a deterministic, single period economic dispatch problem with no transmission constraints or as complex as a stochastic, security-constrained, combined unit-commitment and multiperiod optimal power flow problem with locational contingency and load-following reserves, ramping costs and constraints, deferrable demands, lossy storage resources and uncertain renewable generation.

MOST is documented separately from the main MATPOWER package in its own manual, the MOST User's Manual.

¹http://pserc.org/

²http://www.pserc.cornell.edu/powerweb/

1.2 License and Terms of Use

Beginning with version 5.1, the code in MATPOWER is distributed under the 3-clause BSD license³ [6]. The full text of the license can be found in the LICENSE file at the top level of the distribution or at https://matpower.org/license/ and reads as follows.

Copyright (c) 1996-2016, Power Systems Engineering Research Center (PSERC) and individual contributors (see AUTHORS file for details). All rights reserved.

Redistribution and use in source and binary forms, with or without modification, are permitted provided that the following conditions are met:

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³Versions 4.0 through 5.0 of MATPOWER were distributed under version 3.0 of the GNU General Public License (GPL) [7] with an exception added to clarify our intention to allow MATPOWER to interface with MATLAB as well as any other MATLAB code or MEX-files a user may have installed, regardless of their licensing terms. The full text of the GPL can be found at https://www.gnu.org/licenses/gpl-3.0.txt. MATPOWER versions prior to version 4 had their own license.

Please note that the MATPOWER case files distributed with MATPOWER are not covered by the BSD license. In most cases, the data has either been included with permission or has been converted from data available from a public source.

1.3 Citing Matpower

We request that publications derived from the use of MATPOWER, or the included data files, explicitly acknowledge that fact by citing the appropriate paper(s) and the software itself.⁴

Papers

All publications derived from the use of MATPOWER, or the included data files, should cite the 2011 MATPOWER paper [1]:

R. D. Zimmerman, C. E. Murillo-Sánchez, and R. J. Thomas, "MATPOWER: Steady-State Operations, Planning and Analysis Tools for Power Systems Research and Education," *Power Systems, IEEE Transactions on*, vol. 26, no. 1, pp. 12–19, Feb. 2011. doi: 10.1109/TPWRS.2010.2051168

Publications derived from the use of the MATPOWER Optimal Scheduling Tool (MOST) should cite the 2013 MOST paper [3], in addition to the 2011 MATPOWER paper [1] above.

C. E. Murillo-Sánchez, R. D. Zimmerman, C. L. Anderson, and R. J. Thomas, "Secure Planning and Operations of Systems with Stochastic Sources, Energy Storage and Active Demand," *Smart Grid, IEEE Transactions on*, vol. 4, no. 4, pp. 2220–2229, Dec. 2013.

doi: 10.1109/TSG.2013.2281001

Work making specific reference to the MATPOWER Interior Point Solver (MIPS) should also cite [8]:

H. Wang, C. E. Murillo-Sánchez, R. D. Zimmerman, and R. J. Thomas, "On Computational Issues of Market-Based Optimal Power Flow," *Power Systems, IEEE Transactions on*, vol. 22, no. 3, pp. 1185–1193, August 2007. doi: 10.1109/TPWRS.2007.901301

Note: Some of the case files included with MATPOWER request the citation of additional publications. This includes the ACTIVSg, PEGASE, and RTE cases. Details are available in the help text at the top of the corresponding case files.

⁴This information is also available in the CITATION file at the top level of the MATPOWER distribution.

Software

For the sake of reproducibility of research results, it is best to cite the specific version of the software used, with the version-specific DOI. For example, for version 7.0 of MATPOWER:

```
R. D. Zimmerman, C. E. Murillo-Sánchez (2019). MATPOWER (Version 7.0) [Software]. Available: https://matpower.org doi: 10.5281/zenodo.3251119
```

To cite the MATPOWER software generally, without reference to a specific version, use the following citation and DOI, with $\langle YEAR \rangle$ replaced by the year of the most recent release:

```
R. D. Zimmerman, C. E. Murillo-Sánchez (< YEAR>). MATPOWER [Software]. Available: https://matpower.org doi: 10.5281/zenodo.3236535
```

A list of versions with release dates and version-specific DOI's can be found via the general DOI at https://doi.org/10.5281/zenodo.3236535.

User's Manuals

The MATPOWER, MIPS and MOST User's Manuals should also be cited explicitly in work that refers to or is derived from their content.⁵ As with the software, the citation and DOI can be version-specific or general, as appropriate. For version 7.0 of the MATPOWER User's Manual, use:

```
R. D. Zimmerman, C. E. Murillo-Sánchez. MATPOWER User's Manual, Version 7.0. 2019. [Online]. Available: https://matpower.org/docs/MATPOWER-manual-7.0.pdf doi: 10.5281/zenodo.3251118
```

For a version non-specific citation, use the following citation and DOI, with $\langle YEAR \rangle$ replaced by the year of the most recent release:

```
R. D. Zimmerman, C. E. Murillo-Sánchez. MATPOWER User's Manual. < YEAR>. [Online]. Available: https://matpower.org/docs/MATPOWER-manual.pdf doi: 10.5281/zenodo.3236519
```

A list of versions of the User's Manual with release dates and version-specific DOI's can be found via the general DOI at https://doi.org/10.5281/zenodo.3236519.

⁵For information on citing the MIPS or MOST User's Manuals, please see the mips/CITATION and most/CITATION files, respectively.

Recommendation

In the interest of facilitating research reproducibility and thereby increasing the value of your MATPOWER-related research publications, we strongly encourage you to also publish, whenever possible, all of the code and data required to generate the results you are publishing. Zenodo/GitHub⁶ and IEEE DataPort⁷ are two of many available options⁸.

1.4 Matpower Development

Following the release of MATPOWER 6.0, the MATPOWER project moved to an open development paradigm, hosted on the MATPOWER GitHub project page:

https://github.com/MATPOWER/matpower

The MATPOWER GitHub project hosts the public Git code repository as well as a public issue tracker for handling bug reports, patches, and other issues and contributions. There are separate GitHub hosted repositories and issue trackers for MATPOWER, MOST, MIPS and the testing framework used by all of them, MP-Test, all available from https://github.com/MATPOWER/.

1.5 Sponsoring the Matpower Project

If you have found MATPOWER to be valuable, please consider supporting the project by becoming a sponsor at https://matpower.org/sponsor.

As a free, open-source tool, MATPOWER development relies on research grants and contracts as well as support from sponsors and donors. The software is and always has been provided free of charge under an open-source license, but its development and support require significant resources. Any contributions from the community or other sponsors free us to focus on that support and the development of valuable new features.

⁶See the GitHub Guide on "Making Your Code Citable" at https://guides.github.com/activities/citable-code/.

⁷https://ieee-dataport.org

⁸See the Registry of Research Data Repositories at https://www.re3data.org.

2 Getting Started

2.1 System Requirements

To use Matpower 7.0 you will need:

- Matlab[®] version 7.3 (R2006b) or later⁹, or
- GNU Octave version 4 or later¹⁰

See Section 2.1 in the MOST User's Manual for any requirements specific to MOST.

For the hardware requirements, please refer to the system requirements for the version of Matlab¹¹ or Octave that you are using. If the Matlab Optimization Toolbox is installed as well, Matpower enables an option to use it to solve optimal power flow problems.

In this manual, references to Matlab usually apply to Octave as well. At the time of writing, none of the optional MEX-based Matpower packages have been built for Octave, but Octave does typically include GLPK.

⁹Matlab is available from The MathWorks, Inc. (https://www.mathworks.com/). Matpower 5 and Matpower 6 required Matlab 7 (R14), Matpower 4 required Matlab 6.5 (R13), Matpower 3.2 required Matlab 6 (R12), Matpower 3.0 required Matlab 5 and Matpower 2.0 and earlier required only Matlab 4. Matlab is a registered trademark of The MathWorks, Inc.

¹⁰GNU Octave [5] is free software, available online at https://www.gnu.org/software/octave/. MATPOWER 5 and MATPOWER 6 required Octave 3.4, MATPOWER 4 required Octave 3.2.

¹¹https://www.mathworks.com/support/sysreq/previous_releases.html

2.2 Getting Matpower

You can either download an official versioned release or you can obtain the current development version, which we also attempt to keep stable enough for everyday use. The development version includes new features and bug fixes added since the last versioned release.

2.2.1 Versioned Releases

Download the ZIP file of the latest official versioned release from the MATPOWER website. ¹² Note: This does include the MATPOWER Extras. ¹³

2.2.2 Current Development Version

There are also two options for obtaining the most recent development version of MATPOWER from the master branch on GitHub. Note: This does *not* include the MATPOWER Extras.

- 1. Clone the Matpower repository from GitHub.
 - From the command line:
 - git clone https://github.com/MATPOWER/matpower.git
 - Or, from the MATPOWER GitHub repository page¹⁴:
 - Click the green Clone or download button, then Open in Desktop.

Use this option if you want to be able to easily update to the current development release, with the latest bug fixes and new features, using a simple git pull command, or if you want to help with testing or development. This requires that you have a Git client¹⁵ (GUI or command-line) installed.

- 2. Download a zip file of the MATPOWER repository from GitHub.
 - Go to the Matpower GitHub repository page.
 - Click the green Clone or download button, then Download ZIP.

¹²https://matpower.org

¹³For details on the MATPOWER Extras, see Appendix E and the corresponding GitHub repository at https://github.com/MATPOWER/matpower-extras.

¹⁴https://github.com/MATPOWER/matpower

¹⁵See https://git-scm.com/downloads for information on downloading Git clients.

Use this option if you need features or fixes introduced since the latest versioned release, but you do not have access to or are not ready to begin using Git (but don't be afraid to give Git a try). ¹⁶

See CONTRIBUTING.md for information on how to get a local copy of your own MATPOWER fork, if you are interesting in contributing your own code or modifications.

2.2.3 Matpower Docker Image

MATPOWER is also available on Docker Hub¹⁷ as the pre-packaged Docker¹⁸ image tagged matpower/matpower-desktop,¹⁹ providing an Ubuntu desktop environment with Octave, MATPOWER, and the MATPOWER Extras pre-installed. See the MATPOWER Docker page²⁰ for more details.

Docker images are provided for both versioned releases and development versions.

¹⁶A good place to start is https://git-scm.com.

¹⁷https://hub.docker.com/

¹⁸https://www.docker.com/

¹⁹https://hub.docker.com/r/matpower/matpower-desktop

²⁰https://github.com/MATPOWER/matpower/blob/master/docker/MATPOWER-Docker.md

2.3 Installation

Installation and use of MATPOWER requires familiarity with the basic operation of MATLAB or Octave. Make sure you follow the installation instructions for the version of MATPOWER you are installing. The process was simplified with an install script following version 6.0.

Step 1: Get a copy of Matpower as described above.

Clone the repository or download and extract the ZIP file of the MAT-POWER distribution and place the resulting directory in the location of your choice and call it anything you like.²¹ We will use <MATPOWER> as a placeholder to denote the path to this directory (the one containing install_matpower.m). The files in <MATPOWER> should not need to be modified, so it is recommended that they be kept separate from your own code.

Step 2: Run the installer.

Open MATLAB or Octave and change to the < MATPOWER> directory.

Run the installer and follow the directions to add the required directories to your MATLAB or Octave path, by typing:

install_matpower

Step 3: That's it. There is no step 3.

But, if you chose not to have the installer run the test suite for you in step 2, you can run it now to verify that MATPOWER is installed and functioning properly, by typing:²²

test_matpower

The result should resemble the following, possibly including extra tests, depending on the availablility of optional packages, solvers and extras.

²¹Do not place MATPOWER's files in a directory named 'matlab' or 'optim' (both case-insensitive), as these can cause MATLAB's built-in ver command to behave strangely in ways that affect MATPOWER.

²²The MOST test suite is run separately by typing test_most. See the MOST User's Manual for details.

```
>> test_matpower
t_nested_struct_copy....ok
t_have_fcn....ok
t_feval_w_path....ok
t_mpoption.....ok
\verb|t_loadcase.....ok|
t_ext2int2ext.....ok
t_jacobian.....ok
t_hessian....ok
t_margcost.....ok
t_totcost.....ok
\verb|t_modcost..........ok|
t_hasPQcap.....ok
t_mplinsolve..........ok (6 of 44 skipped)
t_mips.....ok
t_mips_pardiso.........ok (60 of 60 skipped)
t_qps_matpower.......ok (288 of 360 skipped)
t_miqps_matpower......ok (240 of 240 skipped)
\texttt{t\_pf}.....ok
t_pf_radial.....ok
\texttt{t\_cpf}.....ok
t_islands....ok
{\tt t\_opf\_model.....ok}
t_opf_model_legacy.....ok
t_opf_default.....ok
t_opf_mips.....ok (278 of 1296 skipped)
t_opf_dc_mips.....ok
t_opf_dc_mips_sc....ok
t_opf_userfcns.....ok
t_{opf}softlims.....ok
t_runopf_w_res.....ok
t_dcline....ok
t_get_losses.....ok
t_load2disp.....ok
{\tt t\_makePTDF}.....ok
t_makeLODF.....ok
t_printpf.....ok
t_vdep_load....ok
t_total_load.....ok
{\tt t\_scale\_load}.....{\tt ok}
t_apply_changes.....ok
\texttt{t\_psse}.....ok
t_off2case....ok
t_auction_mips.....ok
t_runmarket.....ok
All tests successful (8242 passed, 872 skipped of 9114)
Elapsed time 40.35 seconds.
```

2.4 Running a Simulation

The primary functionality of MATPOWER is to solve power flow and optimal power flow (OPF) problems. This involves (1) preparing the input data defining the all of the relevant power system parameters, (2) invoking the function to run the simulation and (3) viewing and accessing the results that are printed to the screen and/or saved in output data structures or files.

2.4.1 Preparing Case Input Data

The input data for the case to be simulated are specified in a set of data matrices packaged as the fields of a MATLAB struct, referred to as a "MATPOWER case" struct and conventionally denoted by the variable mpc. This struct is typically defined in a case file, either a function M-file whose return value is the mpc struct or a MAT-file that defines a variable named mpc when loaded²³. The main simulation routines, whose names begin with run (e.g. runpf, runopf), accept either a file name or a MATPOWER case struct as an input.

Use loadcase to load the data from a case file into a struct if you want to make modifications to the data before passing it to the simulation.

```
>> mpc = loadcase(casefilename);
```

See also savecase for writing a MATPOWER case struct to a case file.

The structure of the MATPOWER case data is described a bit further in Section 3.1 and the full details are documented in Appendix B and can be accessed at any time via the command help caseformat. The MATPOWER distribution also includes many example case files listed in Table D-18.

2.4.2 Solving the Case

The solver is invoked by calling one of the main simulation functions, such as runpf or runopf, passing in a case file name or a case struct as the first argument. For example, to run a simple Newton power flow with default options on the 9-bus system defined in case9.m, at the MATLAB prompt, type:

```
>> runpf('case9');
```

²³This describes version 2 of the MATPOWER case format, which is used internally and is the default. The version 1 format, now deprecated, but still accessible via the loadcase and savecase functions, defines the data matrices as individual variables rather than fields of a struct, and some do not include all of the columns defined in version 2.

If, on the other hand, you wanted to load the 30-bus system data from case30.m, increase its real power demand at bus 2 to 30 MW, then run an AC optimal power flow with default options, this could be accomplished as follows:

```
>> define_constants;
>> mpc = loadcase('case30');
>> mpc.bus(2, PD) = 30;
>> runopf(mpc);
```

The define_constants in the first line is simply a convenience script that defines a number of variables to serve as named column indices for the data matrices. In this example, it allows us to access the "real power demand" column of the bus matrix using the name PD without having to remember that it is the 3rd column.

Other top-level simulation functions are available for running DC versions of power flow and OPF, for running an OPF with the option for MATPOWER to shut down (decommit) expensive generators, etc. These functions are listed in Table D-2 in Appendix D.

2.4.3 Accessing the Results

By default, the results of the simulation are pretty-printed to the screen, displaying a system summary, bus data, branch data and, for the OPF, binding constraint information. The bus data includes the voltage, angle and total generation and load at each bus. It also includes nodal prices in the case of the OPF. The branch data shows the flows and losses in each branch. These pretty-printed results can be saved to a file by providing a filename as the optional 3rd argument to the simulation function.

The solution is also stored in a results struct available as an optional return value from the simulation functions. This results struct is a superset of the MATPOWER case struct mpc, with additional columns added to some of the existing data fields and additional fields. The following example shows how simple it is, after running a DC OPF on the 118-bus system in case118.m, to access the final objective function value, the real power output of generator 6 and the power flow in branch 51.

```
>> define_constants;
>> results = rundcopf('case118');
>> final_objective = results.f;
>> gen6_output = results.gen(6, PG);
>> branch51_flow = results.branch(51, PF);
```

Full documentation for the content of the results struct can be found in Sections 4.4 and 6.6.

2.4.4 Setting Options

MATPOWER has many options for selecting among the available solution algorithms, controlling the behavior of the algorithms and determining the details of the pretty-printed output. These options are passed to the simulation routines as a MATPOWER options struct. The fields of the struct have names that can be used to set the corresponding value via the mpoption function. Calling mpoption with no arguments returns the default options struct, the struct used if none is explicitly supplied. Calling it with a set of name and value pairs modifies the default vector.

For example, the following code runs a power flow on the 300-bus example in case300.m using the fast-decoupled (XB version) algorithm, with verbose printing of the algorithm progress, but suppressing all of the pretty-printed output.

```
>> mpopt = mpoption('pf.alg', 'FDXB', 'verbose', 2, 'out.all', 0);
>> results = runpf('case300', mpopt);
```

To modify an existing options struct, for example, to turn the verbose option off and re-run with the remaining options unchanged, simply pass the existing options as the first argument to mpoption.

```
>> mpopt = mpoption(mpopt, 'verbose', 0);
>> results = runpf('case300', mpopt);
```

See Appendix C or type:

```
>> help mpoption
```

for more information on Matpower's options.

2.5 Documentation

There are four primary sources of documentation for MATPOWER. The first is this manual, which gives an overview of MATPOWER's capabilities and structure and describes the modeling and formulations behind the code. It can be found in your MATPOWER distribution at <math representation of the latest version is always available at: https://matpower.org/docs/MATPOWER-manual.pdf.

Secondly, the MOST User's Manual describes MOST and its problem formulation, features, data formats and options. It is located at <mar/>MATPOWER>/docs/MOST-manual.pdf in your MATPOWER distribution and the latest version is also available online at: https://matpower.org/docs/MOST-manual.pdf.

The MATPOWER Online Function Reference is the third source of documentation, allowing you to view not only the help text, but also the code itself for each MATPOWER function directly from a web browser. It is available online at: https://matpower.org/docs/ref

Last, but certainly not least, is the built-in help command. As with MATLAB's built-in functions and toolbox routines, you can type help followed by the name of a command or M-file to get help on that particular function. Nearly all of MATPOWER'S M-files have such documentation and this should be considered the main reference for the calling options for each individual function. See Appendix D for a list of MATPOWER functions.

As an example, the help for runopf looks like:

```
>> help runopf
 RUNOPF Runs an optimal power flow.
    [RESULTS, SUCCESS] = RUNOPF(CASEDATA, MPOPT, FNAME, SOLVEDCASE)
    Runs an optimal power flow (AC OPF by default), optionally returning
    a RESULTS struct and SUCCESS flag.
    Inputs (all are optional):
        CASEDATA: either a MATPOWER case struct or a string containing
            the name of the file with the case data (default is 'case9')
            (see also CASEFORMAT and LOADCASE)
       MPOPT : MATPOWER options struct to override default options
            can be used to specify the solution algorithm, output options
            termination tolerances, and more (see also MPOPTION).
       FNAME: name of a file to which the pretty-printed output will
            be appended
        SOLVEDCASE: name of file to which the solved case will be saved
            in MATPOWER case format (M-file will be assumed unless the
            specified name ends with '.mat')
    Outputs (all are optional):
        RESULTS: results struct, with the following fields:
            (all fields from the input MATPOWER case, i.e. bus, branch,
                gen, etc., but with solved voltages, power flows, etc.)
            order - info used in external <-> internal data conversion
            et - elapsed time in seconds
            success - success flag, 1 = succeeded, 0 = failed
            (additional OPF fields, see OPF for details)
        SUCCESS: the success flag can additionally be returned as
            a second output argument
    Calling syntax options:
       results = runopf;
       results = runopf(casedata);
        results = runopf(casedata, mpopt);
        results = runopf(casedata, mpopt, fname);
        results = runopf(casedata, mpopt, fname, solvedcase);
        [results, success] = runopf(...);
        Alternatively, for compatibility with previous versions of MATPOWER,
        some of the results can be returned as individual output arguments:
        [baseMVA, bus, gen, gencost, branch, f, success, et] = runopf(...);
   Example:
       results = runopf('case30');
                                      24
   See also RUNDCOPF, RUNUOPF.
```

3 Modeling

MATPOWER employs all of the standard steady-state models typically used for power flow analysis. The AC models are described first, then the simplified DC models. Internally, the magnitudes of all values are expressed in per unit and angles of complex quantities are expressed in radians. Internally, all off-line generators and branches are removed before forming the models used to solve the power flow, continuation power flow or optimal power flow problem, and all buses are numbered consecutively, beginning at 1. Conversions to and from this internal indexing is done by the functions ext2int and int2ext. The notation in this section, as well as Sections 4 and 6, is based on this internal numbering, with all generators and branches assumed to be in-service. Due to the strengths of the MATLAB programming language in handling matrices and vectors, the models and equations are presented here in matrix and vector form.

3.1 Data Formats

The data files used by MATPOWER are MATLAB M-files or MAT-files which define and return a single MATLAB struct. The M-file format is plain text that can be edited using any standard text editor. The fields of the struct are baseMVA, bus, branch, gen and optionally gencost, where baseMVA is a scalar and the rest are matrices. In the matrices, each row corresponds to a single bus, branch, or generator. The columns are similar to the columns in the standard IEEE CDF and PTI formats. The number of rows in bus, branch and gen are n_b , n_l and n_g , respectively. If present, gencost has either n_g or $2n_g$ rows, depending on whether it includes costs for reactive power or just real power. Full details of the MATPOWER case format are documented in Appendix B and can be accessed from the MATLAB command line by typing help caseformat.

3.2 Branches

All transmission lines²⁴, transformers and phase shifters are modeled with a common branch model, consisting of a standard π transmission line model, with series impedance $z_s = r_s + jx_s$ and total charging susceptance b_c , in series with an ideal phase shifting transformer. The transformer, whose tap ratio has magnitude τ and

²⁴This does not include DC transmission lines. For more information the handling of DC transmission lines in MATPOWER, see Section 7.6.3.

phase shift angle $\theta_{\rm shift}$, is located at the *from* end of the branch, as shown in Figure 3-1. The parameters r_s , x_s , b_c , τ and $\theta_{\rm shift}$ are specified directly in columns BR_R (3), BR_X (4), BR_B (5), TAP (9) and SHIFT (10), respectively, of the corresponding row of the branch matrix.²⁵

The complex current injections i_f and i_t at the from and to ends of the branch, respectively, can be expressed in terms of the 2×2 branch admittance matrix Y_{br} and the respective terminal voltages v_f and v_t

$$\begin{bmatrix} i_f \\ i_t \end{bmatrix} = Y_{br} \begin{bmatrix} v_f \\ v_t \end{bmatrix}. \tag{3.1}$$

With the series admittance element in the π model denoted by $y_s = 1/z_s$, the branch admittance matrix can be written

$$Y_{br} = \begin{bmatrix} \left(y_s + j\frac{b_c}{2}\right) \frac{1}{\tau^2} & -y_s \frac{1}{\tau e^{-j\theta_{\text{shift}}}} \\ -y_s \frac{1}{\tau e^{j\theta_{\text{shift}}}} & y_s + j\frac{b_c}{2} \end{bmatrix}. \tag{3.2}$$

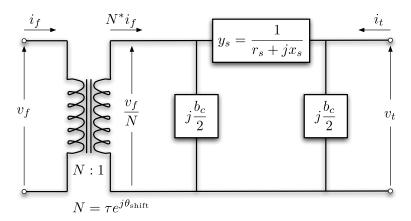


Figure 3-1: Branch Model

If the four elements of this matrix for branch i are labeled as follows:

$$Y_{br}^{i} = \begin{bmatrix} y_{ff}^{i} & y_{ft}^{i} \\ y_{tf}^{i} & y_{tt}^{i} \end{bmatrix}$$

$$(3.3)$$

then four $n_l \times 1$ vectors Y_{ff} , Y_{ft} , Y_{tf} and Y_{tt} can be constructed, where the *i*-th element of each comes from the corresponding element of Y_{br}^i . Furthermore, the $n_l \times n_b$ sparse

²⁵A value of zero in the TAP column indicates that the branch is a transmission line and not a transformer, i.e. mathematically equivalent to a transformer with tap ratio set to 1.

connection matrices C_f and C_t used in building the system admittance matrices can be defined as follows. The $(i, j)^{\text{th}}$ element of C_f and the $(i, k)^{\text{th}}$ element of C_t are equal to 1 for each branch i, where branch i connects from bus j to bus k. All other elements of C_f and C_t are zero.

3.3 Generators

A generator is modeled as a complex power injection at a specific bus. For generator i, the injection is

$$s_q^i = p_q^i + jq_q^i. (3.4)$$

Let $S_g = P_g + jQ_g$ be the $n_g \times 1$ vector of these generator injections. The MW and MVAr equivalents (before conversion to p.u.) of p_g^i and q_g^i are specified in columns PG (2) and QG (3), respectively of row i of the gen matrix. A sparse $n_b \times n_g$ generator connection matrix C_g can be defined such that its $(i,j)^{\text{th}}$ element is 1 if generator j is located at bus i and 0 otherwise. The $n_b \times 1$ vector of all bus injections from generators can then be expressed as

$$S_{g,\text{bus}} = C_g \cdot S_g. \tag{3.5}$$

A generator with a negative injection can also be used to model a dispatchable load.

3.4 Loads

Constant power loads are modeled as a specified quantity of real and reactive power consumed at a bus. For bus i, the load is

$$s_d^i = p_d^i + jq_d^i (3.6)$$

and $S_d = P_d + jQ_d$ denotes the $n_b \times 1$ vector of complex loads at all buses. The MW and MVAr equivalents (before conversion to p.u.) of p_d^i and q_d^i are specified in columns PD (3) and QD (4), respectively of row i of the bus matrix. These fields can also take on negative quantities to represent fixed (e.g. distributed) generation.

Constant impedance and constant current loads are not implemented directly, but the constant impedance portions can be modeled as a shunt element described below. Dispatchable loads are modeled as negative generators and appear as negative values in S_q .

3.5 Shunt Elements

A shunt connected element such as a capacitor or inductor is modeled as a fixed impedance to ground at a bus. The admittance of the shunt element at bus i is given as

$$y_{sh}^{i} = g_{sh}^{i} + jb_{sh}^{i} (3.7)$$

and $Y_{sh} = G_{sh} + jB_{sh}$ denotes the $n_b \times 1$ vector of shunt admittances at all buses. The parameters g_{sh}^i and b_{sh}^i are specified in columns GS (5) and BS (6), respectively, of row i of the bus matrix as equivalent MW (consumed) and MVAr (injected) at a nominal voltage magnitude of 1.0 p.u and angle of zero.

A shunt element can also be used to model a constant impedance load and, though correctly, MATPOWER does not currently report these quantities as "load".

3.6 Network Equations

For a network with n_b buses, all constant impedance elements of the model are incorporated into a complex $n_b \times n_b$ bus admittance matrix Y_{bus} that relates the complex nodal current injections I_{bus} to the complex node voltages V:

$$I_{\text{bus}} = Y_{\text{bus}}V. \tag{3.8}$$

Similarly, for a network with n_l branches, the $n_l \times n_b$ system branch admittance matrices Y_f and Y_t relate the bus voltages to the $n_l \times 1$ vectors I_f and I_t of branch currents at the *from* and *to* ends of all branches, respectively:

$$I_f = Y_f V \tag{3.9}$$

$$I_t = Y_t V. (3.10)$$

If $[\cdot]$ is used to denote an operator that takes an $n \times 1$ vector and creates the corresponding $n \times n$ diagonal matrix with the vector elements on the diagonal, these system admittance matrices can be formed as follows:

$$Y_f = [Y_{ff}] C_f + [Y_{ft}] C_t (3.11)$$

$$Y_t = [Y_{tf}] C_f + [Y_{tt}] C_t (3.12)$$

$$Y_{\text{bus}} = C_f^{\mathsf{T}} Y_f + C_t^{\mathsf{T}} Y_t + [Y_{sh}].$$
 (3.13)

The current injections of (3.8)–(3.10) can be used to compute the corresponding

complex power injections as functions of the complex bus voltages V:

$$S_{\text{bus}}(V) = [V] I_{\text{bus}}^* = [V] Y_{\text{bus}}^* V^*$$
 (3.14)

$$S_f(V) = [C_f V] I_f^* = [C_f V] Y_f^* V^*$$
(3.15)

$$S_t(V) = [C_t V] I_t^* = [C_t V] Y_t^* V^*.$$
(3.16)

The nodal bus injections are then matched to the injections from loads and generators to form the AC nodal power balance equations, expressed as a function of the complex bus voltages and generator injections in complex matrix form as

$$g_S(V, S_g) = S_{\text{bus}}(V) + S_d - C_g S_g = 0.$$
 (3.17)

3.7 DC Modeling

The DC formulation [20] is based on the same parameters, but with the following three additional simplifying assumptions.

• Branches can be considered lossless. In particular, branch resistances r_s and charging capacitances b_c are negligible:

$$y_s = \frac{1}{r_s + jx_s} \approx \frac{1}{jx_s}, \quad b_c \approx 0.$$
 (3.18)

• All bus voltage magnitudes are close to 1 p.u.

$$v_i \approx e^{j\theta_i}. (3.19)$$

• Voltage angle differences across branches are small enough that

$$\sin(\theta_f - \theta_t - \theta_{\text{shift}}) \approx \theta_f - \theta_t - \theta_{\text{shift}}.$$
 (3.20)

Substituting the first set of assumptions regarding branch parameters from (3.18), the branch admittance matrix in (3.2) approximates to

$$Y_{br} \approx \frac{1}{jx_s} \begin{bmatrix} \frac{1}{\tau^2} & -\frac{1}{\tau e^{-j\theta_{\text{shift}}}} \\ -\frac{1}{\tau e^{j\theta_{\text{shift}}}} & 1 \end{bmatrix}.$$
 (3.21)

Combining this and the second assumption with (3.1) yields the following approximation for i_f :

$$i_f \approx \frac{1}{jx_s} \left(\frac{1}{\tau^2} e^{j\theta_f} - \frac{1}{\tau e^{-j\theta_{\text{shift}}}} e^{j\theta_t} \right)$$

$$= \frac{1}{jx_s \tau} \left(\frac{1}{\tau} e^{j\theta_f} - e^{j(\theta_t + \theta_{\text{shift}})} \right). \tag{3.22}$$

The approximate real power flow is then derived as follows, first applying (3.19) and (3.22), then extracting the real part and applying (3.20).

$$p_{f} = \Re \left\{ s_{f} \right\}$$

$$= \Re \left\{ v_{f} \cdot i_{f}^{*} \right\}$$

$$\approx \Re \left\{ e^{j\theta_{f}} \cdot \frac{j}{x_{s}\tau} \left(\frac{1}{\tau} e^{-j\theta_{f}} - e^{-j(\theta_{t} + \theta_{\text{shift}})} \right) \right\}$$

$$= \Re \left\{ \frac{j}{x_{s}\tau} \left(\frac{1}{\tau} - e^{j(\theta_{f} - \theta_{t} - \theta_{\text{shift}})} \right) \right\}$$

$$= \Re \left\{ \frac{1}{x_{s}\tau} \left[\sin(\theta_{f} - \theta_{t} - \theta_{\text{shift}}) + j \left(\frac{1}{\tau} - \cos(\theta_{f} - \theta_{t} - \theta_{\text{shift}}) \right) \right] \right\}$$

$$\approx \frac{1}{x_{s}\tau} (\theta_{f} - \theta_{t} - \theta_{\text{shift}})$$
(3.23)

As expected, given the lossless assumption, a similar derivation for the power injection at the to end of the line leads to leads to $p_t = -p_f$.

The relationship between the real power flows and voltage angles for an individual branch i can then be summarized as

$$\begin{bmatrix} p_f \\ p_t \end{bmatrix} = B_{br}^i \begin{bmatrix} \theta_f \\ \theta_t \end{bmatrix} + P_{\text{shift}}^i$$
 (3.24)

where

$$\begin{split} B_{br}^i &= b_i \left[\begin{array}{cc} 1 & -1 \\ -1 & 1 \end{array} \right], \\ P_{\text{shift}}^i &= \theta_{\text{shift}}^i b_i \left[\begin{array}{c} -1 \\ 1 \end{array} \right] \end{split}$$

and b_i is defined in terms of the series reactance x_s^i and tap ratio τ^i for branch i as

$$b_i = \frac{1}{x_s^i \tau^i}.$$

For a shunt element at bus i, the amount of complex power consumed is

$$s_{sh}^{i} = v_{i}(y_{sh}^{i}v_{i})^{*}$$

$$\approx e^{j\theta_{i}}(g_{sh}^{i} - jb_{sh}^{i})e^{-j\theta_{i}}$$

$$= g_{sh}^{i} - jb_{sh}^{i}.$$
(3.25)

So the vector of real power consumed by shunt elements at all buses can be approximated by

$$P_{sh} \approx G_{sh}. (3.26)$$

With a DC model, the linear network equations relate real power to bus voltage angles, versus complex currents to complex bus voltages in the AC case. Let the $n_l \times 1$ vector B_{ff} be constructed similar to Y_{ff} , where the *i*-th element is b_i and let $P_{f,\text{shift}}$ be the $n_l \times 1$ vector whose *i*-th element is equal to $-\theta_{\text{shift}}^i b_i$. Then the nodal real power injections can be expressed as a linear function of Θ , the $n_b \times 1$ vector of bus voltage angles

$$P_{\text{bus}}(\Theta) = B_{\text{bus}}\Theta + P_{\text{bus,shift}} \tag{3.27}$$

where

$$P_{\text{bus,shift}} = (C_f - C_t)^{\mathsf{T}} P_{f,\text{shift}}. \tag{3.28}$$

Similarly, the branch flows at the *from* ends of each branch are linear functions of the bus voltage angles

$$P_f(\Theta) = B_f \Theta + P_{f,\text{shift}} \tag{3.29}$$

and, due to the lossless assumption, the flows at the to ends are given by $P_t = -P_f$. The construction of the system B matrices is analogous to the system Y matrices for the AC model:

$$B_f = [B_{ff}] (C_f - C_t) (3.30)$$

$$B_{\text{bus}} = \left(C_f - C_t\right)^{\mathsf{T}} B_f. \tag{3.31}$$

The DC nodal power balance equations for the system can be expressed in matrix form as

$$g_P(\Theta, P_g) = B_{\text{bus}}\Theta + P_{\text{bus,shift}} + P_d + G_{sh} - C_g P_g = 0$$
(3.32)

4 Power Flow

The standard power flow or loadflow problem involves solving for the set of voltages and flows in a network corresponding to a specified pattern of load and generation. Matpower includes solvers for both AC and DC power flow problems, both of which involve solving a set of equations of the form

$$g(x) = 0, (4.1)$$

constructed by expressing a subset of the nodal power balance equations as functions of unknown voltage quantities.

All of Matpower's solvers exploit the sparsity of the problem and, except for Gauss-Seidel, scale well to very large systems. Currently, none of them include any automatic updating of transformer taps or other techniques to attempt to satisfy typical optimal power flow constraints, such as generator, voltage or branch flow limits.

4.1 AC Power Flow

In MATPOWER, by convention, a single generator bus is typically chosen as a reference bus to serve the roles of both a voltage angle reference and a real power slack. The voltage angle at the reference bus has a known value, but the real power generation at the slack bus is taken as unknown to avoid overspecifying the problem. The remaining generator buses are typically classified as PV buses, with the values of voltage magnitude and generator real power injection given. These are specified in the VG (6) and PG (3) columns of the gen matrix, respectively. Since the loads P_d and Q_d are also given, all non-generator buses are classified as PQ buses, with real and reactive injections fully specified, taken from the PD (3) and QD (4) columns of the bus matrix. Let \mathcal{I}_{ref} , \mathcal{I}_{PV} and \mathcal{I}_{PQ} denote the sets of bus indices of the reference bus, PV buses and PQ buses, respectively. The bus type classification is specified in the MATPOWER case file in the BUS_TYPE column (2) of the bus matrix. Any isolated buses must be identified as such in this column as well.

In the traditional formulation of the AC power flow problem, the power balance equation in (3.17) is split into its real and reactive components, expressed as functions of the voltage angles Θ and magnitudes V_m and generator injections P_g and Q_g , where the load injections are assumed constant and given:

$$g_P(\Theta, V_m, P_g) = P_{\text{bus}}(\Theta, V_m) + P_d - C_g P_g = 0$$
 (4.2)

$$g_Q(\Theta, V_m, Q_q) = Q_{\text{bus}}(\Theta, V_m) + Q_d - C_q Q_q = 0.$$
 (4.3)

For the AC power flow problem, the function g(x) from (4.1) is formed by taking the left-hand side of the real power balance equations (4.2) for all non-slack buses and the reactive power balance equations (4.3) for all PQ buses and plugging in the reference angle, the loads and the known generator injections and voltage magnitudes:

$$g(x) = \begin{bmatrix} g_P^{\{i\}}(\Theta, V_m, P_g) \\ g_Q^{\{j\}}(\Theta, V_m, Q_g) \end{bmatrix} \quad \forall i \in \mathcal{I}_{PV} \cup \mathcal{I}_{PQ} \\ \forall j \in \mathcal{I}_{PQ}.$$
 (4.4)

The vector x consists of the remaining unknown voltage quantities, namely the voltage angles at all non-reference buses and the voltage magnitudes at PQ buses:

$$x = \begin{bmatrix} \theta_{\{i\}} \\ v_m^{\{j\}} \end{bmatrix} \quad \forall i \notin \mathcal{I}_{ref}$$

$$\forall j \in \mathcal{I}_{PQ}.$$
 (4.5)

This yields a system of nonlinear equations with $n_{pv} + 2n_{pq}$ equations and unknowns, where n_{pv} and n_{pq} are the number of PV and PQ buses, respectively. After solving for x, the remaining real power balance equation can be used to compute the generator real power injection at the slack bus. Similarly, the remaining $n_{pv} + 1$ reactive power balance equations yield the generator reactive power injections.

MATPOWER includes four different algorithms for solving the general AC power flow problem.²⁶ The default solver is based on a standard Newton's method [10] using a polar form and a full Jacobian updated at each iteration. Each Newton step involves computing the mismatch g(x), forming the Jacobian based on the sensitivities of these mismatches to changes in x and solving for an updated value of x by factorizing this Jacobian. This method is described in detail in many textbooks. As of version 7, MATPOWER also includes variants of the Newton method using nodal current balance equations and/or cartesian representations for voltage as described in [11].

Also included are solvers based on variations of the fast-decoupled method [12], specifically, the XB and BX methods described in [13]. These solvers greatly reduce the amount of computation per iteration, by updating the voltage magnitudes and angles separately based on constant approximate Jacobians which are factored only once at the beginning of the solution process. These per-iteration savings, however, come at the cost of more iterations.

The fourth algorithm is the standard Gauss-Seidel method from Glimm and Stagg [14]. It has numerous disadvantages relative to the Newton method and is included primarily for academic interest.

²⁶Three more that are specific to radial networks typical of distribution systems are described in Section 4.3.

By default, the AC power flow solvers simply solve the problem described above, ignoring any generator limits, branch flow limits, voltage magnitude limits, etc. However, there is an option (pf.enforce_q.lims) that allows for the generator reactive power limits to be respected at the expense of the voltage setpoint. This is done in a rather brute force fashion by adding an outer loop around the AC power flow solution. If any generator has a violated reactive power limit, its reactive injection is fixed at the limit, the corresponding bus is converted to a PQ bus and the power flow is solved again. This procedure is repeated until there are no more violations. Note that this option is based solely on the QMAX and QMIN parameters for the generator, from columns 4 and 5 of the gen matrix, and does not take into account the trapezoidal generator capability curves described in Section 6.4.3 and specifed in columns PC1-QC2MAX (11-16). Note also that this option affects generators even if the bus they are attached to is already of type PQ.

In all, there are 4 different formulations implemented for the Newton power flow, described briefly below, with the details found in reference [11].

4.1.1 Cartesian vs. Polar Coordinates for Voltage

Another variation of the standard AC power flow problem represents the bus voltages in cartesian, rather than polar, coordinates. That is, instead of Θ and V_m , the vector x of unknowns includes the real and imaginary parts of the complex voltage, denoted respectively by U and W, where V = U + jW. The active and reactive power balance equations from (4.2)–(4.3) are then expressed as functions of U and W.

$$g_P(U, W, P_g) = P_{\text{bus}}(U, W) + P_d - C_g P_g = 0$$
 (4.6)

$$g_Q(U, W, Q_g) = Q_{\text{bus}}(U, W) + Q_d - C_g Q_g = 0.$$
 (4.7)

The power flow problem of (4.4) becomes

$$g(x) = \begin{bmatrix} g_P^{\{i\}}(U, W, P_g) \\ g_Q^{\{j\}}(U, W, Q_g) \end{bmatrix} \quad \forall i \in \mathcal{I}_{PV} \cup \mathcal{I}_{PQ} \\ \forall j \in \mathcal{I}_{PQ}.$$
 (4.8)

where

$$x = \begin{bmatrix} u_{\{i\}} \\ w^{\{j\}} \end{bmatrix} \qquad \forall i \in \mathcal{I}_{PQ}$$

$$\forall j \notin \mathcal{I}_{ref}.$$
 (4.9)

For PV buses, since the voltage magnitude is constant, Δu can be expressed as a function of Δw , allowing u to be eliminated for PV buses by modifying the Jacobian terms corresponding to g_P and w. This results once again in a system of nonlinear equations with $n_{pv} + 2n_{pq}$ equations and unknowns.

In MATPOWER setting the pf.v_cartesian option to 1 (0 by default) selects the cartesian representation for voltages when running a Newton-Raphson AC power flow, that is, when the pf.alg option is set to 'NR'.²⁷

4.1.2 Current vs. Power for Nodal Balance Equations

Another variation of the standard AC power flow problem uses current balance equations in place of the power balance equations in (4.4) and (4.8). If we let M and N represent the real and imaginary parts, respectively, of the current, we can express the current balance functions for the polar form as

$$g_M(\Theta, V_m, P_g, Q_g) = \Re\{I_{\text{bus}}(\Theta, V_m) + [V^*]^{-1}(S_d - C_g S_g)^*\} = 0$$
(4.10)

$$g_N(\Theta, V_m, P_g, Q_g) = \Im\{I_{\text{bus}}(\Theta, V_m) + [V^*]^{-1}(S_d - C_g S_g)^*\} = 0$$
 (4.11)

and for the cartesian form as

$$g_M(U, W, P_q, Q_q) = \Re\{I_{\text{bus}}(U, W) + [V^*]^{-1}(S_d - C_q S_q)^*\} = 0$$
 (4.12)

$$g_N(U, W, P_q, Q_q) = \Im\{I_{\text{bus}}(U, W) + [V^*]^{-1}(S_d - C_q S_q)^*\} = 0$$
 (4.13)

where $S_d = P_d + jQ_d$, $S_g = P_g + jQ_g$ and $[V^*]^{-1}$ is a diagonal matrix whose *i*-th diagonal entry is $1/v_i^*$, that is $\frac{1}{v_m^i}e^{j\theta_i}$ or $1/(u_i - jw_i)$.

In order to handle PV buses, g(x) is expanded to include the full set of mismatch equations for all PV and PQ buses, and the reactive power injections at PV buses are added to x as independent variables, resulting in a system of nonlinear equations with $2n_{pv} + 2n_{pq}$ equations and unknowns.

In MATPOWER setting the pf.current_balance option to 1 (0 by default) selects the current balance version when running a Newton-Raphson AC power flow, that is, when the pf.alg option is set to 'NR'. This option can be used in conjunction with either the polar or cartesian representation of bus voltages.²⁸

4.2 DC Power Flow

For the DC power flow problem [20], the vector x consists of the set of voltage angles at non-reference buses

$$x = [\theta_{\{i\}}], \quad \forall i \notin \mathcal{I}_{\text{ref}}$$
 (4.14)

²⁷Setting the pf.alg option to 'NR-SC' is also a shortcut for selecting this version based on cartesian voltage representation.

²⁸Setting the pf.alg option to 'NR-IP' or 'NR-IC' are also shortcuts for selecting the polar and cartesian versions, respectively, of the current balance formulation.

and
$$(4.1)$$
 takes the form

$$B_{dc}x - P_{dc} = 0 (4.15)$$

where B_{dc} is the $(n_b - 1) \times (n_b - 1)$ matrix obtained by simply eliminating from B_{bus} the row and column corresponding to the slack bus and reference angle, respectively. Given that the generator injections P_g are specified at all but the slack bus, P_{dc} can be formed directly from the non-slack rows of the last four terms of (3.32).

The voltage angles in x are computed by a direct solution of the set of linear equations. The branch flows and slack bus generator injection are then calculated directly from the bus voltage angles via (3.29) and the appropriate row in (3.32), respectively.

4.3 Distribution Power Flow

Distribution systems are different from transmission systems in a number of respects, such as the x_s/r_s branch ratio, magnitudes of x_s and r_s and most importantly the typically radial structure. Due to these differences, a number of power flow solution methods have been developed to account for the specific nature of distribution systems and most widely used are the backward/forward sweep methods [15, 16]. MATPOWER includes an additional three AC power flow methods that are specific to radial networks.

4.3.1 Radial Power Flow

When solving radial distribution networks it is practical to number branches with numbers that are equal to the receiving bus numbers. An example is given in Figure 4-1, where branches are drawn black. Furthermore, the oriented branch ordering [17] offers a possibility for fast and efficient backward/forward sweeps. All branches are always oriented from the sending bus to the receiving bus and the only requirement is that the sending bus number should be smaller than the receiving bus number. This means that i < k for branch i-k. The indices of the sending nodes of branches are stored in vector F such that $i = f_k$.

As usual, the supply bus (slack bus) is given index 1, meaning that branch indices should go from 2 to n_b which is the number of buses in the network. Introducing a fictitious branch with index 1 and zero impedance, given with dashed black line in Figure 4-1, the number of branches n_l becomes equal to the number of buses n_b .

For the example of Figure 4-1 vector F is the following

$$F = \begin{bmatrix} 0 & 1 & 2 & 3 & 3 & 2 & 4 \end{bmatrix}^\mathsf{T},$$

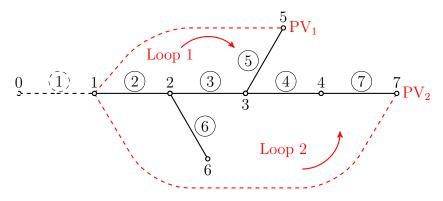


Figure 4-1: Oriented Ordering

and it offers an easy way to follow the path between any bus and bus 0. If we consider bus 4, the path to the slack bus consists of following branches: branch 4, since considered bus is their receiving bus; branch 3, since $f_4 = 3$; branch 2, since $f_3 = 2$; and branch 1, since $f_2 = 1$.

The representation of branch k, connecting buses i and k, is given in Figure 4-2, where it is modeled with its serial impedance z_s^k . At both ends there are load demands s_d^i and s_d^k , and shunt admittances y_d^i and y_d^k comprised of admittances due capacitance of all lines and shunt elements connected to buses i and k

$$y_d^k = j\left(\sum b_{\text{lines}}^k + \sum b_{\text{shunt}}^k\right).$$

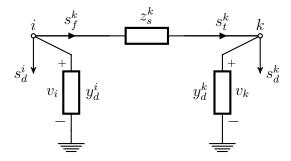


Figure 4-2: Branch Representation: branch k between buses i (sending) and k (receiving) and load demand and shunt admittances at both buses

4.3.2 Current Summation Method

The voltage calculation procedure with the Current Summation Method is performed in 5 steps as follows [15, 16].

- 1. Set all voltages to 1 p.u. (flat start). Set iteration count $\nu = 1$.
- 2. Set branch current flow equal to the sum of current of the demand at receiving end (s_d^k) and the current drawn in the admittance (y_d^k) connected to bus k

$$j_b^k = \left(\frac{s_d^k}{v_k}\right)^* + y_d^k \cdot v_k, \quad k = 1, 2, \dots, n_b.$$
 (4.16)

3. Backward sweep: Perform current summation, starting from the branch with the biggest index and heading towards the branch whose index is equal to 1. The current of branch k is added to the current of the branch whose index is equal to $i = f_k$.

$$j_{h \text{ new}}^i = j_h^i + j_h^k, \quad k = n_l, n_l - 1, \dots, 2$$
 (4.17)

4. Forward sweep: The receiving end bus voltages are calculated with known branch currents and sending bus voltages.

$$v_k = v_i - z_s^k \cdot j_b^k, \quad k = 2, 3, \dots, n_l$$
 (4.18)

5. Compare voltages in iteration ν with the corresponding ones from iteration $\nu-1$. If the maximum difference in magnitude is less than the specified tolerance

$$\max_{i=1\dots n_k} \left\{ \left| v_i^{\nu} - v_i^{\nu-1} \right| \right\} < \varepsilon \tag{4.19}$$

the procedure is finished. Otherwise go to step 2.

4.3.3 Power Summation Method

The voltage calculation procedure with the Power Summation Method is performed in 5 steps as follows [17].

1. Set all voltages to 1 p.u. (flat start). Set iteration count $\nu = 1$.

2. Set receiving end branch flow equal to the sum of the demand at receiving end (s_d^k) and the power drawn in the admittance (y_d^k) connected to bus k

$$s_t^k = s_d^k + \frac{(y_d^k)^*}{v_k^2}, \quad k = 1, 2, \dots, n_b.$$
 (4.20)

3. Backward sweep: Calculate sending end branch power flows as a sum of receiving end branch power flows and branch losses via (4.21). Perform power summation, starting from the branch with the biggest index and heading towards the branch whose index is equal to 1. The sending power of branch k is added to the receiving power of the branch whose index is equal to $i = f_k$ as in (4.22).

$$s_f^k = s_t^k + z_s^k \cdot \left| \frac{s_t^k}{v_k} \right|^2 \qquad k = n_l, n_l - 1, \dots, 2$$
 (4.21)

$$s_{t,\text{new}}^i = s_t^i + s_f^k$$
 $k = n_l, n_l - 1, \dots, 2$ (4.22)

4. Forward sweep: The receiving end bus voltages are calculated with known sending powers and voltages.

$$v_k = v_i - z_s^k \cdot \left(\frac{s_f^k}{v_i}\right)^* \quad k = 2, 3, \dots, n_l$$
 (4.23)

5. Compare voltages in iteration ν with the corresponding ones from iteration $\nu - 1$, using (4.19). If the maximum difference in magnitude is less than the specified tolerance the procedure is finished. Otherwise go to step 2.

4.3.4 Admittance Summation Method

For each node, besides the known admittance y_d^k , we define the admittance y_e^k as the driving point admittance of the part of the network fed by node k, including shunt admittance y_d^k . We also define an equivalent current generator j_e^k for the part of the network fed by node k. The current of this generator consists of all load currents fed by node k. The process of calculation of bus voltages with the admittance summation method consists of the following 5 steps [18].

1. Set all voltages to 1 p.u. (flat start). Set iteration count $\nu=1$. Set initial values

$$y_e^k = y_d^k, \quad k = 1, 2, \dots, n_b.$$
 (4.24)

2. For each node k, calculate equivalent admittance y_e^k . Perform admittance summation, starting from the branch with the biggest index and heading towards the branch whose index is equal to 1. The driving point admittance of branch kis added to the driving point admittance of the branch whose index is equal to $i = f_k$ as in (4.26).

$$d_b^k = \frac{1}{1 + z_a^k \cdot y_a^k} \qquad k = n_l, n_l - 1, \dots, 2$$
 (4.25)

$$y_{e,\text{new}}^i = y_e^i + d_b^k \cdot y_e^k$$
 $k = n_l, n_l - 1, \dots, 2$ (4.26)

3. Backward sweep: For each node k calculate equivalent current generator j_e^k , first set it equal to load current j_d^k and perform current summation over equivalent admittances using factor d_b^k as in (4.25).

$$j_e^k = j_d^k = \left(\frac{s_d^k}{v_k}\right)^* \qquad k = n_l, n_l - 1, \dots, 2 \qquad (4.27)$$

$$k_{\text{new}} = j_e^i + d_b^k \cdot j_e^k \qquad k = n_l, n_l - 1, \dots, 2 \qquad (4.28)$$

$$j_{e,\text{new}}^i = j_e^i + d_b^k \cdot j_e^k$$
 $k = n_l, n_l - 1, \dots, 2$ (4.28)

4. Forward sweep: The receiving end bus voltages are calculated with known equivalent current generators and sending bus voltages.

$$v_k = d_b^k \cdot (v_i - z_s^k \cdot j_e^k) \quad k = 2, 3, \dots, n_l$$
 (4.29)

5. Compare voltages in iteration ν with the corresponding ones from iteration $\nu - 1$, using (4.19). If the maximum difference in magnitude is less than the specified tolerance the procedure is finished. Otherwise go to step 3.

Handling PV Buses 4.3.5

The methods explained in the previous three subsections are applicable to radial networks without loops and PV buses. These methods can be used to solve the power flow problem in weakly meshed networks if a compensation procedure based on The venin equivalent impedance matrix is added [15,16]. In [17] a voltage correction procedure is added to the process.

The list of branches is expanded by a set of n_{pv} fictitious links, corresponding to the PV nodes. Each of these links starts at the slack bus and ends at a corresponding PV bus, thus forming a loop in the network. A fictitious link going to bus k is represented by a voltage generator with a voltage magnitude equal to the specified voltage magnitude at bus k. Its phase angle is equal to the calculated phase angle at bus k.

A loop impedance matrix Z_l is formed for the loops made by the fictitious links and it has the following properties

- Element z_l^{mm} is equal to the sum of branch impedances of all branches related to loop m,
- Element z_l^{mk} is equal to the sum of branch impedances of mutual branches of loops m and k.

As an illustration, in Figure 4-1 there a two PV generators at buses 5 and 7. The fictitious links and loops orientation are drawn in red. The Thevenin matrix for this case is

$$Z_{l} = \begin{bmatrix} z_{b}^{2} + z_{b}^{3} + z_{b}^{5} & z_{b}^{2} + z_{b}^{3} \\ z_{b}^{2} + z_{b}^{3} & z_{b}^{2} + z_{b}^{3} + z_{b}^{4} + z_{b}^{7} \end{bmatrix}.$$

First column elements are equal to PV bus voltages when the current injection at bus 5 is 1 p.u. and $v_1 = 0$. Bus voltages can be calculated with the current summation method in a single iteration. By repeating the procedure for bus 7 one can calculate the elements of second column.

By breaking all links the network becomes radial [16] and the three backward/forward sweep methods are applicable. Since all links are fictitious, only the injected reactive power at their receiving bus m is determined [17] by the following equation

$$\Delta q_{pv}^m = \Delta d_{pv}^m \frac{v_m^2}{\Re \left\{ v_m \right\}},\tag{4.30}$$

which is practically an increment in reactive power injection of the corresponding PV generator for the current iteration.

The incremental changes of the imaginary part of PV generator current Δd_{pv}^m can be obtained by solving the matrix equation

$$\Im \{Z_l\} \cdot \Delta D_{pv} = \Delta E_{pv},\tag{4.31}$$

where

$$\Delta e_{pv}^{m} = \left(\frac{v_g^{m}}{|v_m|} - 1\right) \cdot \Re\{v_m\}, \quad m = 1, 2, \dots, n_{pv}.$$
(4.32)

In order to ensure 90° phase difference between voltage and current at PV generators in [19] it was suggested to calculate the real part of PV generator current

$$\Delta c_{pv}^{m} = \Delta d_{pv}^{m} \frac{\Im \left\{ v_{m} \right\}}{\Re \left\{ v_{m} \right\}}.$$
(4.33)

In such a way the PV generator will inject purely reactive power, as it is supposed to do. Its active power is added before as a negative load.

Before proceeding with the next iteration, the bus voltage corrections are calculated. In order to do that, the radial network is solved by applying incremental current changes $\Delta I_{pv} = \Delta C_{pv} + j\Delta D_{pv}$ at the PV buses as excitations and setting $v_1 = 0$. After the backward/forward sweep is performed with the current summation method, the voltage corrections at all buses are known. They are added to the latest voltages in order to obtain the new bus voltages, which are used in the next iteration [17].

4.4 runpf

In MATPOWER, a power flow is executed by calling runpf with a case struct or case file name as the first argument (casedata). In addition to printing output to the screen, which it does by default, runpf optionally returns the solution in a results struct.

```
>> results = runpf(casedata);
```

The results struct is a superset of the input MATPOWER case struct mpc, with some additional fields as well as additional columns in some of the existing data fields. The solution values are stored as shown in Table 4-1.

Additional optional input arguments can be used to set options (mpopt) and provide file names for saving the pretty printed output (fname) or the solved case data (solvedcase).

```
>> results = runpf(casedata, mpopt, fname, solvedcase);
```

The options that control the power flow simulation are listed in Table 4-2 and those controlling the output printed to the screen in Table 4-3.

By default, runpf solves an AC power flow problem using a standard Newton's method solver. To run a DC power flow, the model option must be set to 'DC'. For convenience, MATPOWER provides a function rundcpf which is simply a wrapper that sets the model option to 'DC' before calling runpf.

Internally, the runpf function does a number of conversions to the problem data before calling the appropriate solver routine for the selected power flow algorithm.

Table 4-1: Power Flow Results

name	description
results.success results.et results.iterations results.order results.bus(:, VM)† results.bus(:, VA)	success flag, 1 = succeeded, 0 = failed computation time required for solution number of iterations required for solution see ext2int help for details on this field bus voltage magnitudes bus voltage angles
results.gen(:, PG) results.gen(:, QG) [†] results.branch(:, PF) results.branch(:, PT) results.branch(:, QF) [†] results.branch(:, QT) [†]	generator real power injections generator reactive power injections real power injected into "from" end of branch real power injected into "to" end of branch reactive power injected into "from" end of branch reactive power injected into "to" end of branch

[†] AC power flow only.

This external-to-internal format conversion is performed by the ext2int function, described in more detail in Section 7.3.1, and includes the elimination of out-of-service equipment and the consecutive renumbering of buses. All computations are done using this internal indexing. When the simulation has completed, the data is converted back to external format by int2ext before the results are printed and returned.

4.5 Linear Shift Factors

The DC power flow model can also be used to compute the sensitivities of branch flows to changes in nodal real power injections, sometimes called injection shift factors (ISF) or generation shift factors [20]. These $n_l \times n_b$ sensitivity matrices, also called power transfer distribution factors or PTDFs, carry an implicit assumption about the slack distribution. If H is used to denote a PTDF matrix, then the element in row i and column j, h_{ij} , represents the change in the real power flow in branch i given a unit increase in the power injected at bus j, with the assumption that the additional unit of power is extracted according to some specified slack distribution:

$$\Delta P_f = H \Delta P_{\text{bus}}.\tag{4.34}$$

This slack distribution can be expressed as an $n_b \times 1$ vector w of non-negative weights whose elements sum to 1. Each element specifies the proportion of the slack taken up at each bus. For the special case of a single slack bus k, w is equal to the

vector e_k . The corresponding PTDF matrix H_k can be constructed by first creating the $n_l \times (n_b - 1)$ matrix

$$\widetilde{H}_k = \widetilde{B}_f \cdot B_{dc}^{-1} \tag{4.35}$$

then inserting a column of zeros at column k. Here \widetilde{B}_f and B_{dc} are obtained from B_f and B_{bus} , respectively, by eliminating their reference bus columns and, in the case of B_{dc} , removing row k corresponding to the slack bus.

The PTDF matrix H_w , corresponding to a general slack distribution w, can be obtained from any other PTDF, such as H_k , by subtracting $H_k \cdot w$ from each column, equivalent to the following simple matrix multiplication:

$$H_w = H_k(I - w \cdot 1^\mathsf{T}). \tag{4.36}$$

These same linear shift factors may also be used to compute sensitivities of branch flows to branch outages, known as line outage distribution factors or LODFs [21]. Given a PTDF matrix H_w , the corresponding $n_l \times n_l$ LODF matrix L can be constructed as follows, where l_{ij} is the element in row i and column j, representing the change in flow in branch i (as a fraction of the initial flow in branch j) for an outage of branch j.

First, let H represent the matrix of sensitivities of branch flows to branch endpoint injections, found by multplying the PTDF matrix by the node-branch incidence matrix:

$$H = H_w (C_f - C_t)^\mathsf{T}. \tag{4.37}$$

Here the individual elements h_{ij} represent the sensitivity of flow in branch i with respect to injections at branch j endpoints, corresponding to a simulated increase in flow in branch j. Then l_{ij} can be expressed as

$$l_{ij} = \begin{cases} \frac{h_{ij}}{1 - h_{jj}} & i \neq j \\ -1 & i = j. \end{cases}$$
 (4.38)

MATPOWER includes functions for computing both the DC PTDF matrix and the corresponding LODF matrix for either a single slack bus k or a general slack distribution vector w. See the help for makePTDF and makeLODF and Sections 9.5.5 and 9.5.4, respectively, for details.

Table 4-2: Power Flow Options

name	default	description
model	'AC'	AC vs. DC modeling for power flow and OPF formulation 'AC' – use AC formulation and corresponding alg options 'DC' – use DC formulation and corresponding alg options
pf.alg	'NR'	AC power flow algorithm: 'NR' - Newtons's method (formulation depends on values of pf.current_balance and pf.v_cartesian options) 'NR-SC' - Newtons's method (power mismatch, cartesian)† 'NR-IP' - Newtons's method (current mismatch, polar)† 'NR-IC' - Newtons's method (current mismatch, cartesian)† 'FDXB' - Fast-Decoupled (XB version) 'FDBX' - Fast-Decouple (BX version) 'GS' - Gauss-Seidel 'PQSUM' - Power Summation (radial networks only) 'ISUM' - Current Summation (radial networks only) 'YSUM' - Admittance Summation (radial networks only)
pf.current_balance [‡]	0	use current, as opposed to power, balance for AC PF, 0 or 1
pf.v_cartesian [‡]	0	use cartesian, as opposed to polar, representation for voltages for AC PF, 0 or 1
pf.tol	10^{-8}	termination tolerance on per unit P and Q dispatch
pf.nr.max_it	10	maximum number of iterations for Newton's method
pf.nr.lin_solver	1 1	linear solver option for mplinsolve for computing Newton update step (see mplinsolve for complete list of all options) '' - default to '\' for small systems, 'LU3' for larger ones
		'\' - built-in backslash operator 'LU' - explicit default LU decomposition and back substitution 'LU3' - 3 output arg form of lu, Gilbert-Peierls algorithm with approximate minimum degree (AMD) reordering 'LU4' - 4 output arg form of lu, UMFPACK solver (same as 'LU')
		'LU5' – 5 output arg form of lu, UMFPACK solver w/row scaling
pf.fd.max_it	30	maximum number of iterations for fast-decoupled method
pf.gs.max_it	1000	maximum number of iterations for Gauss-Seidel method
pf.radial.max_it	20	maximum number of iterations for radial power flow methods
pf.radial.vcorr	0	perform voltage correction procedure in distribution power flow 0 – do not perform voltage correction
pf.enforce_q_lims	0	1 – perform voltage correction enforce gen reactive power limits at expense of $ V_m $ 0 – do not enforce limits 1 – enforce limits, simultaneous bus type conversion 2 – enforce limits, one-at-a-time bus type conversion

 $^{^\}dagger$ Automatically sets/overrides pf.current_balance and pf.v_cartesian options with corresponding values. ‡ Only relevant for Newton power flow solver. 45

Table 4-3: Power Flow Output Options

name	default	description
verbose	1	amount of progress info to be printed
		0 – print no progress info
		1 – print a little progress info
		2 – print a lot of progress info
		3 – print all progress info
out.all	-1	controls pretty-printing of results
		-1 – individual flags control what is printed
		$0-{ m do}\ not\ { m print\ anything}^\dagger$
		$1 - \text{print everything}^{\dagger}$
out.sys_sum	1	print system summary (0 or 1)
out.area_sum	0	print area summaries (0 or 1)
out.bus	1	print bus detail, includes per bus gen info (0 or 1)
out.branch	1	print branch detail (0 or 1)
out.gen	0	print generator detail (0 or 1)
out.force	0	print results even if success flag $= 0 (0 \text{ or } 1)$
${\tt out.suppress_detail}$	-1	suppress all output but system summary
		-1 – suppress details for large systems (> 500 buses)
		0 – do not suppress any output specified by other flags
		1 – suppress all output except system summary section †

 $^{^{\}dagger}$ Overrides individual flags, but (in the case of out.suppress_detail) not out.all = 1.

5 Continuation Power Flow

Continuation methods or branch tracing methods are used to trace a curve given an initial point on the curve. These are also called predictor-corrector methods since they involve the prediction of the next solution point and correcting the prediction to get the next point on the curve.

Consider a system of n nonlinear equations $g(x) = 0, x \in \mathbb{R}^n$. By adding a continuation parameter λ and one more equation to the system, x can be traced by varying λ . The resulting system $f(x,\lambda) = 0$ has n+1 dimensions. The additional equation is a parameterized equation which identifies the location of the current solution with respect to the previous or next solution.

The continuation process can be diagrammatically shown by (5.1).

$$(x^j, \lambda^j) \xrightarrow{Predictor} (\hat{x}^{j+1}, \hat{\lambda}^{j+1}) \xrightarrow{Corrector} (x^{j+1}, \lambda^{j+1})$$
 (5.1)

where, (x^j, λ^j) represents the current solution at step j, $(\hat{x}^{j+1}, \hat{\lambda}^{j+1})$ is the predicted solution for the next step, and (x^{j+1}, λ^{j+1}) is the next solution on the curve.

Continuation methods are employed in power systems to determine steady state stability limits [22] in what is called a continuation power flow²⁹. The limit is determined from a nose curve where the nose represents the maximum power transfer that the system can handle given a power transfer schedule. To determine the steady state loading limit, the basic power flow equations

$$g(x) = \begin{bmatrix} P(x) - P^{inj} \\ Q(x) - Q^{inj} \end{bmatrix} = 0, \tag{5.2}$$

are restructured as

$$f(x,\lambda) = g(x) - \lambda b = 0 \tag{5.3}$$

where $x \equiv (\Theta, V_m)$ and b is a vector of power transfer given by

$$b = \begin{bmatrix} P_{target}^{inj} - P_{base}^{inj} \\ Q_{target}^{inj} - Q_{base}^{inj} \end{bmatrix}.$$
 (5.4)

The effects of the variation of loading or generation can be investigated using the continuation method by composing the b vector appropriately.

²⁹Thanks to Shrirang Abhyankar, Rui Bo, and Alexander Flueck for contributions to MAT-POWER's continuation power flow feature.

5.1 Parameterization

The values of (x, λ) along the solution curve can parameterized in a number of ways [23, 24]. Parameterization is a mathematical way of identifying each solution so that the *next* solution or *previous* solution can be quantified. MATPOWER includes three parameterization scheme options to quantify this relationship, detailed below, where σ is the continuation step size parameter.

• Natural parameterization simply uses λ directly as the parameter, so the new λ is simply the previous value plus the step size.

$$p^{j}(x,\lambda) = \lambda - \lambda^{j} - \sigma^{j} = 0 \tag{5.5}$$

• Arc length parameterization results in the following relationship, where the step size is equal to the 2-norm of the distance from one solution to the next.

$$p^{j}(x,\lambda) = \sum_{i} (x_{i} - x_{i}^{j})^{2} + (\lambda - \lambda^{j})^{2} - (\sigma^{j})^{2} = 0$$
 (5.6)

• Pseudo arc length parameterization [26] is MATPOWER's default parameterization scheme, where the next point (x, λ) on the solution curve is constrained to lie in the hyperplane running through the predicted solution $(\hat{x}^{j+1}, \hat{\lambda}^{j+1})$ orthogonal to the tangent line from the previous corrected solution (x^j, λ^j) . This relationship can be quantified by the function

$$p^{j}(x,\lambda) = \left(\begin{bmatrix} x \\ \lambda \end{bmatrix} - \begin{bmatrix} x^{j} \\ \lambda^{j} \end{bmatrix} \right)^{\mathsf{T}} \bar{z}^{j} - \sigma^{j} = 0, \tag{5.7}$$

where \bar{z}^j is the normalized tangent vector at (x^j, λ^j) and σ^j is the continuation step size parameter.

5.2 Predictor

The predictor is used to produce an estimate for the next solution. The better the prediction, the faster is the convergence to the solution point. MATPOWER uses a tangent predictor for estimating the curve to the next solution. The tangent vector $z^j = \begin{bmatrix} dx & d\lambda \end{bmatrix}_j^\mathsf{T}$ at the current solution (x^j, λ^j) is found by solving the linear system

$$\begin{bmatrix} f_x & f_{\lambda} \\ p_x^{j-1} & p_{\lambda}^{j-1} \end{bmatrix} z^j = \begin{bmatrix} 0 \\ 1 \end{bmatrix}.$$
 (5.8)

The matrix on the left-hand side is simply the standard power flow Jacobian with an additional column and row added. The extra column f_{λ} is simply the negative of the power transfer vector b and the extra row, required to make the system non-singular and define the magnitude of z^{j} , is the derivative of the parameterization function at the previous solution point p^{j-1} .

The resulting tangent vector is then normalized

$$\bar{z}^j = \frac{z^j}{||z^j||_2} \tag{5.9}$$

and used to compute the predicted approximation $(\hat{x}^{j+1}, \hat{\lambda}^{j+1})$ to the next solution (x^{j+1}, λ^{j+1}) using

$$\begin{bmatrix} \hat{x}^{j+1} \\ \hat{\lambda}^{j+1} \end{bmatrix} = \begin{bmatrix} x^j \\ \lambda^j \end{bmatrix} + \sigma^j \bar{z}^j, \tag{5.10}$$

where σ^{j} is the continuation step size.

5.3 Corrector

The corrector stage finds the next solution (x^{j+1}, λ^{j+1}) by correcting the approximation estimated by the predictor $(\hat{x}^{j+1}, \hat{\lambda}^{j+1})$. Newton's method is used to find the next solution by solving the n+1 dimensional system in (5.11), where one of (5.5)–(5.7) has been added as an additional constraint to the parameterized power flow equations of (5.3).

$$\begin{bmatrix} f(x,\lambda) \\ p^{j}(x,\lambda) \end{bmatrix} = 0 \tag{5.11}$$

5.4 Step Length Control

Step length control is a key element affecting the computational efficiency of a continuation method. It affects the continuation method with two issues: (1) speed – how fast the corrector converges to a specified accuracy, and (2) robustness – whether the corrector converges to a true solution given a predicted point. Matpower's continuation power flow can optionally use adaptive steps, where the step size σ is adjusted by a scaling factor α within the limits specified.

$$\sigma^{j+1} = \alpha^j \sigma^j, \qquad \sigma_{\min} \le \sigma^{j+1} \le \sigma_{\max}$$
 (5.12)

This scaling factor α^j for step j is limited to a maximum of 2 and is calculated from an error estimation between the predicted and corrected solutions γ^j as follows,

$$\alpha^{j} = 1 + \beta_{\text{cpf}} \left(\frac{\epsilon_{\text{cpf}}}{\gamma^{j}} - 1 \right), \qquad \alpha^{j} \le 2,$$
 (5.13)

where $\beta_{\rm cpf}$ is a damping factor, $\epsilon_{\rm cpf}$ is a specified tolerance, and γ^j is given by

$$\gamma^{j} = \left\| \left(x^{j+1}, \lambda^{j+1} \right) - \left(\hat{x}^{j+1}, \hat{\lambda}^{j+1} \right) \right\|_{\infty}. \tag{5.14}$$

5.5 Event Detection and Location

A continuation power flow *event* is triggered when the value of one of the elements of an event function changes sign from one continuation step to the next. The event occurs at the point where the corresponding value of the event function passes through zero. Matpower provides event functions to detect the location at which the continuation power flow reaches the following:

- a specified target λ value
- the nose point
- the end of a full trace
- a generator reactive power limit
- a generator active power limit
- a bus voltage magnitude limit
- a branch flow limit

Each event function is registered with an event name, a flag indicating whether or not the location of the event should be pinpointed, and if so, to within what tolerance. For events that are to be located, when an event interval is detected, that is, when an element of the event function value changes sign, MATPOWER adjusts the continuation step size via a False Position or Regula Falsi method until it locates the point of the zero-crossing to within the specified tolerance.

The detection of an event zero, or even an event interval, can be used to trigger further actions. The CPF callback function capability was extended in MATPOWER 6.x to include the ability to handle events by including information about any event intervals or zeros detected. For example, MATPOWER includes a callback that fixes the reactive output of a generator and converts its bus from PV to PQ when the corresponding event function indicates that its reactive power limit has been reached. Another responds to the detection of the nose point by signaling the termination of the continuation. In fact, continuation power flow termination for nose point, target

lambda or full trace modes are all based on CPF callback functions in conjunction with event detection.

While MATPOWER does include a mechanism for supplying user defined callback functions, it does not yet have a corresponding mechanism for user specified event functions.

5.6 runcpf

In MATPOWER, a continuation power flow is executed by calling runcpf with two MATPOWER cases (case structs or case file names) as the first two arguments, basecasedata and targetcasedata, respectively. The first contains the base loading/generation profile while the second contains the target loading/generation profile. In addition to printing output to the screen, which it does by default, runcpf optionally returns the solution in a results struct.

```
>> results = runcpf(basecasedata, targetcasedata);
```

Additional optional input arguments can be used to set options (mpopt) and provide file names for saving the pretty printed output (fname) or the solved case data (solvedcase).

```
>> results = runcpf(basecasedata, targetcasedata, mpopt, fname, solvedcase);
```

The results struct is a superset of the input MATPOWER case struct mpc, with some additional fields as well as additional columns in some of the existing data fields. In addition to the solution values included in the results for a simple power flow, shown in Table 4-1 in Section 4.4, the following additional continuation power flow solution values are stored in the cpf field as shown in Table 5-1.

The options that control the continuation power flow simulation are listed in Table 5-2. All the power flow options for Newton's method (tolerance, maximum iterations) and for controlling the output on the screen (see Tables 4-2 and 4-3) are also available with the continuation power flow.

Table 5-1: Continuation Power Flow Results

name	description
results.cpf	CPF output struct whose content depends on any user callback functions, where default contains fields:
$done_msg$	string with message describing cause of continuation termination
events(eidx)	a structure array of size n_e , where n_e is the number of events located, with fields:
k	continuation step number at which event was located
name	name of event
idx	index(es) of critical elements in corresponding event function, e.g. index of generator reaching VAr limit
msg	descriptive text detailing the event
iterations	n_{steps} , number of continuation steps performed
lam	$1 \times n$ vector of λ values from correction steps [†]
lam_hat	$1 \times n$ vector of λ values from prediction steps [†]
max_lam	maximum value of λ found in results.cpf.lam
steps	$1 \times n$ vector of step sizes for each continuation step performed [†]
V	$n_b \times n$ matrix of complex bus voltages from correction steps [†]
V_hat	$n_b \times n$ matrix of complex bus voltages from prediction steps [†]

[†] n is one more than the number of continuation steps, i.e. $n_{\text{steps}} + 1$, so the first element corresponds to the starting point.

Table 5-2: Continuation Power Flow Options

$\begin{array}{cccccccccccccccccccccccccccccccccccc$	of parameterization satural rc length seeds arc length seeds arc length sines stopping criterion $'$ — stop when nose point is reached $'$ — trace full nose curve — stop upon reaching target λ value λ_{stop} e gen active power limits so not enforce limits inforce limits e gen reactive power limits at expense of V_m to not enforce limits inforce limits
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	rc length secudo arc length dines stopping criterion \cdot — stop when nose point is reached \cdot — trace full nose curve — stop upon reaching target λ value λ_{stop} and e gen active power limits do not enforce limits at expense of V_m do not enforce limits and not enforce limits at expense of V_m do not enforce limits and not enforce limits
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	seudo arc length ines stopping criterion \cdot — stop when nose point is reached \cdot — trace full nose curve — stop upon reaching target λ value λ_{stop} e gen active power limits no not enforce limits inforce limits at expense of V_m to not enforce limits inforce limits
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	ines stopping criterion ' — stop when nose point is reached ' — trace full nose curve — stop upon reaching target λ value λ_{stop} e gen active power limits to not enforce limits force limits e gen reactive power limits at expense of V_m to not enforce limits be bus voltage magnitude limits force limits force limits for not enforce limits for not enforce limits for not enforce limits
$\begin{array}{c} \text{'NOSE} \\ \text{'FULL} \\ \lambda_{\text{stop}} \\ \text{cpf.enforce_p_lims} & 0 & \text{enforce} \\ 0 & -d \\ 1 - e \\ \text{cpf.enforce_q_lims} & 0 & \text{enforce} \\ 0 - d \\ 1 - e \\ \text{cpf.enforce_v_lims} & 0 & \text{enforce} \\ \text{cpf.enforce_v_lims} & 0 & \text{enforce} \\ 0 - d \\ 1 - e \\ \text{cpf.enforce_flow_lims} & 0 & \text{enforce} \\ 0 - d \\ 1 - e \\ \end{array}$	'— stop when nose point is reached '— trace full nose curve — stop upon reaching target λ value λ_{stop} e gen active power limits to not enforce limits force limits e gen reactive power limits at expense of V_m to not enforce limits force limits to not enforce limits to not enforce limits force limits to not enforce limits force limits
$\begin{array}{c} \text{ 'FULL} \\ \lambda_{\text{stop}} \\ \text{ cpf.enforce_p_lims} & 0 & \text{enforce} \\ 0 & -d \\ 1 - e \\ \text{cpf.enforce_q_lims} & 0 & \text{enforce} \\ 0 - d \\ 1 - e \\ \text{cpf.enforce_v_lims} & 0 & \text{enforce} \\ 0 - d \\ 1 - e \\ \text{cpf.enforce_flow_lims} & 0 & \text{enforce} \\ 0 - d \\ 1 - e \\ \end{array}$	The trace full nose curve — stop upon reaching target λ value $\lambda_{\rm stop}$ eigen active power limits to not enforce limits eigen reactive power limits at expense of V_m to not enforce limits inforce limits eigen voltage magnitude limits to not enforce limits inforce limits
$\begin{array}{c} \lambda_{\rm stop} \\ -1 \\ -1 \\ -1 \\ -1 \\ -1 \\ -1 \\ -1 \\ -$	- stop upon reaching target λ value λ_{stop} e gen active power limits to not enforce limits e gen reactive power limits at expense of V_m to not enforce limits inforce limits e bus voltage magnitude limits to not enforce limits inforce limits
<pre>cpf.enforce_p_lims</pre>	e gen active power limits no not enforce limits nforce limits e gen reactive power limits at expense of V_m to not enforce limits nforce limits e bus voltage magnitude limits to not enforce limits nforce limits
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	to not enforce limits inforce limits at expense of V_m to not enforce limits inforce limits
$\begin{array}{c} 1-e\\ \text{cpf.enforce_q_lims} & 0 & \text{enforce}\\ 0-d\\ 1-e\\ \text{cpf.enforce_v_lims} & 0 & \text{enforce}\\ 0-d\\ 1-e\\ \text{cpf.enforce_flow_lims} & 0 & \text{enforce}\\ 0-d\\ 1-e\\ \end{array}$	inforce limits v gen reactive power limits at expense of V_m to not enforce limits inforce limits v bus voltage magnitude limits v to v on v enforce limits v inforce limits v inforce limits v inforce limits
$\begin{array}{ccccc} {\rm cpf.enforce_q_lims} & 0 & {\rm enforce} \\ & 0 - {\rm d} \\ & 1 - {\rm e} \\ {\rm cpf.enforce_v_lims} & 0 & {\rm enforce} \\ & 0 - {\rm d} \\ & 1 - {\rm e} \\ {\rm cpf.enforce_flow_lims} & 0 & {\rm enforce} \\ & 0 - {\rm d} \\ & 1 - {\rm e} \\ \end{array}$	e gen reactive power limits at expense of V_m to not enforce limits inforce limits but but voltage magnitude limits to not enforce limits inforce limits
$\begin{array}{c} 0 - d \\ 1 - e \\ \text{cpf.enforce_v_lims} & 0 & \text{enforce} \\ 0 - d \\ 1 - e \\ \text{cpf.enforce_flow_lims} & 0 & \text{enforce} \\ 0 - d \\ 1 - e \\ \end{array}$	to not enforce limits enforce limits to not enforce limits to not enforce limits enforce limits
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	nforce limits bus voltage magnitude limits o not enforce limits nforce limits
$\begin{array}{cccc} {\tt cpf.enforce_v_lims} & 0 & {\tt enforce} \\ & 0 - {\tt d} \\ & 1 - {\tt e} \\ {\tt cpf.enforce_flow_lims} & 0 & {\tt enforce} \\ & 0 - {\tt d} \\ & 1 - {\tt e} \end{array}$	e bus voltage magnitude limits to not enforce limits inforce limits
$\begin{array}{c} 0 - \mathrm{d} \\ 1 - \mathrm{e} \\ \mathrm{cpf.enforce_flow_lims} & 0 & \mathrm{enforce} \\ 0 - \mathrm{d} \\ 1 - \mathrm{e} \end{array}$	o not enforce limits nforce limits
$\begin{array}{ccc} & & 1-\mathrm{e} \\ \mathrm{cpf.enforce_flow_lims} & 0 & & \mathrm{enforce} \\ & 0-\mathrm{d} \\ 1-\mathrm{e} \end{array}$	nforce limits
$\begin{array}{ccc} \texttt{cpf.enforce_flow_lims} & 0 & \texttt{enforce} \\ & 0 - \texttt{d} \\ & 1 - \texttt{e} \end{array}$	
0-d $1-e$	branch MVA flow limits
1 — e	
	o not enforce limits
	nforce limits
cpf.step 0.05 default	value for continuation power flow step size σ
	ım allowed step size, σ_{\min}
	um allowed step size, $\sigma_{\rm max}$
	adaptive step size feature
0 — a	daptive step size disabled
	daptive step size enabled
	ng factor $\beta_{\rm cpf}$ from (5.13) for adaptive step sizing
	$ce \epsilon_{cpf}$ from (5.13) for adaptive step sizing
	ce for target lambda detection
_	ce for nose point detection (p.u.)
=	ce for generator active power limit detection
(MW)	1
	ce for generator reactive power limit detection
(MVAı	
	ce for bus voltage magnitude limit detection (p.u)
1	ce for branch flow limit detection (MVA)
1	plotting of nose curve
	o not plot nose curve
	lot when completed
_	lot incrementally at each iteration
	ame as 2, with pause at each iteration
	of bus whose voltage is to be plotted
	or cell array of strings with names of user callback
function	

 $^{^{\}dagger}$ See help cpf_default_callback for details.

5.6.1 CPF Callback Functions

MATPOWER's continuation power flow provides a callback mechanism to give the user access to the iteration process for executing custom code at each iteration, for example, to implement custom incremental plotting of a PV nose curve or to handle a detected event. This callback mechanism is used internally to handle default plotting functionality as well as to handle the standard CPF events. The $cpf_default_callback$ function, for example, is used to collect the λ and V results from each predictor and corrector iteration and optionally plot the PV nose curve.

The prototype for a CPF callback function is

and the input and output arguments are described in Tables 5-3 through 5-5 and in the help for cpf_default_callback. Each registered CPF callback function is called in three different contexts, distinguished by the value of the first argument k as follows:

- 1. initial called with k = 0, without results input/output arguments, after base power flow, before first CPF step.
- 2. iterations called with k > 0, without results input/output arguments, at each iteration, after predictor-corrector step
- 3. final called with k < 0, with results input/output arguments, after exiting predictor-corrector loop, inputs identical to last iteration call, except k negated

The user can define their own callback functions which take the same form and are called in the same contexts as cpf_default_callback. User callback functions are specified via the MATPOWER option cpf.user_callback. This option can be a string containing the name of the callback function, or a struct with the following fields, where all but the first are optional:

- fcn string with name of callback function
- ullet priority numerical value specifying callback priority, 30 default = 20
- args arbitrary value (any type) passed to the callback as cb_args each time it is invoked

Multiple user callbacks can be registered by assigning a cell array of such strings and/or structs to the cpf.user_callback option.

³⁰See cpf_register_callback for details.

Table 5-3: Continuation Power Flow Callback Input Arguments

name	description
k	continuation step iteration count
nx	next CPF state, corresponding to proposed next step
cx	current CPF state, corresponding to most recent successful step
px	previous CPF state, corresponding to last successful step prior to cx
done	struct, with flag to indicate CPF termination and reason, with fields:
.flag	termination flag, $1 \to \text{terminate}, 0 \to \text{continue}$
.msg	string containing reason for termination
rollback	scalar flag to indicate that the current step should be rolled back and
	retried with a different step size, etc.
evnts	struct array listing any events detected for this step [‡]
cb_data	struct containing static data [§] , with the following fields (all based on internal
	indexing):
$.{\tt mpc_base}$	Matpower case struct of base case
$.{\tt mpc_target}$	Matpower case struct of target case
.Sbusb	handle of function returning $n_b \times 1$ vector of complex base case bus injec-
	tions in p.u. and derivatives w.r.t. $ V $
.Sbust	handle of function returning $n_b \times 1$ vector of complex target case bus
	injections in p.u. and derivatives w.r.t. $ V $
.Ybus	bus admittance matrix
.Yf	branch admittance matrix, "from" end of branches
.Yt	branch admittance matrix, "to" end of branches
.pv	list of indices of PV buses
.pq	list of indices of PQ buses
.ref	list of indices of reference buses
$.\mathtt{idx_pmax}$	vector of gen indices of gens at PMAX
$.{ t mpopt}$	Matpower options struct
${\sf cb_args}$	arbitrary data structure containing user-provided callback arguments
results	initial value of output struct to be assigned to cpf field of results struct returned by runcpf

 $^{^*}$ See Table 5-5 for details of the CPF state.

See cpf_detect_events for details of the evnts field.

Flease note that any callback that modifies the underlying problem is responsible to update the contents of cb_data accordingly. E.g. converting a bus from PV to PQ requires updates to mpc_base, mpc_target, Sbusb, Sbust, pv, pq, and possibly ref. So, cb_data should only be thought of as static for a fixed base and target case pair.

Table 5-4: Continuation Power Flow Callback Output Arguments

name	description	
All are updated versions of the corresponding input arguments, see Table 5-3 for more details.		
nx	next CPF state, user state (cb field) should be updated here if rollback	
	is false	
cx	current CPF state, may contain updated this_step or this_parm values	
	to be used if rollback is true	
done	callback may have requested termination and set the msg field	
rollback	callback can request a rollback step, even if it was not indicated by an event function †	
evnts	msg field for a given event may be updated [‡]	
cb_data	this data should only be modified if the underlying problem has been changed (e.g. generator limit reached), in which case it should always	
	be followed by a step of zero length, i.e. set nx.this_step to 0^{\S}	
results	updated version of results input argument to be assigned to cpf field of results struct returned by runcpf	

See Table 5-5 for details of the CPF state.

Table 5-5: Continuation Power Flow State

-	
name	description
lam_hat	value of λ after predictor step
V_hat	vector of complex bus voltages after predictor step
lam	value of λ after corrector step
V	vector of complex bus voltages after corrector step
Z	normalized tangent predictor, \bar{z}
$default_step$	default step size
$default_parm$	default parameterization*
this_step	step size for this step only
${ t this_parm}$	parameterization* for this step only
step	current step size
parm	current parameterization*
events	event log
cb	user state for callbacks [†] , a user-defined struct containing any information
	a callback function would like to pass from one invokation to the next
ef	cell array of event function values

 $^{^{\}ast}$ Corresponding to the cpf.parameterization option in Table 5-2. † Replaces cb_state from Matpower 5.

[†] In this case, the callback should also modify the step size or parameterization to be used for the re-try, by setting the this_step or this_parm fields in cx.

‡ See cpf_detect_events for details of the evnts field.

§ It is the job of any callback modifying cb_data to ensure that all data in cb_data is kept consistent.

5.6.2 CPF Example

The following is an example of running a continuation power flow using a version of the 9-bus system, looking at increasing all loads by a factor of 2.5. This example plots the nose curve shown in Figure 5-1.

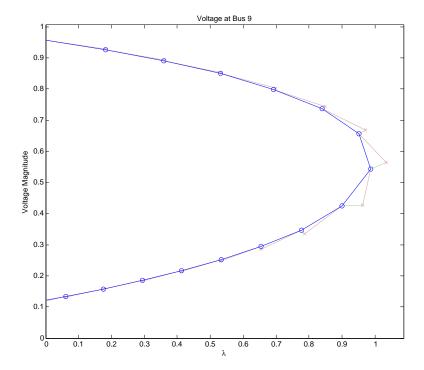


Figure 5-1: Nose Curve of Voltage Magnitude at Bus 9

This should result in something like the following output to the screen.

```
MATPOWER Version 7.0, 20-Jun-2019 -- AC Continuation Power Flow
step
      0 :
                              lambda = 0.000, 4 Newton steps
      1 : stepsize = 0.2
                              lambda = 0.181 2 corrector Newton steps
step
                              lambda = 0.359 2 corrector Newton steps
step
      2 : stepsize = 0.2
                              lambda = 0.530 2 corrector Newton steps
      3 : stepsize = 0.2
step
step 4 : stepsize = 0.2
                              lambda = 0.693 3 corrector Newton steps
step 5 : stepsize = 0.2
                              lambda = 0.839 3 corrector Newton steps
      6 : stepsize = 0.2
                              lambda = 0.952 3 corrector Newton steps
step
                              lambda = 0.988 3 corrector Newton steps
step 7 : stepsize = 0.2
                              lambda = 0.899 3 corrector Newton steps
step 8 : stepsize = 0.2
step 9 : stepsize = 0.2
                              lambda = 0.776 3 corrector Newton steps
                               lambda = 0.654 3 corrector Newton steps
step 10 : stepsize = 0.2
step 11 : stepsize = 0.2
                              lambda = 0.533 2 corrector Newton steps
step 12 : stepsize = 0.2
                              lambda = 0.413 2 corrector Newton steps
step 13 : stepsize = 0.2
                              lambda = 0.294 2 corrector Newton steps
                               lambda = 0.176 2 corrector Newton steps
step 14 : stepsize = 0.2
step 15 : stepsize = 0.2
                              lambda = 0.060 2 corrector Newton steps
step 16a : stepsize = 0.2
                               lambda = -0.054
                                               2 corrector Newton steps ^ ROLLBACK
step 16 : stepsize = 0.0604
                              lambda = 0.000
                                               3 corrector Newton steps
CPF TERMINATION: Traced full continuation curve in 16 continuation steps
```

The results of the continuation power flow are then found in the cpf field of the returned results struct.

```
>> results.cpf
ans =

    V_hat: [9x17 double]
    lam_hat: [1x17 double]
        V: [9x17 double]
        lam: [1x17 double]
        steps: [1x17 double]
        iterations: 16
        max_lam: 0.9876
        events: [1x1 struct]
        done_msg: 'Traced full continuation curve in 16 continuation steps'
```

6 Optimal Power Flow

MATPOWER includes code to solve both AC and DC versions of the optimal power flow problem. The standard version of each takes the following form:

$$\min_{x} f(x) \tag{6.1}$$

subject to

$$g(x) = 0 (6.2)$$

$$h(x) < 0 \tag{6.3}$$

$$x_{\min} \le x \le x_{\max} \quad . \tag{6.4}$$

In both cases, the objective function f(x) consists of the polynomial cost of generator injections, the equality constraints g(x) are the power balance equations, the inequality constraints h(x) are the branch flow limits, and the x_{\min} and x_{\max} bounds include reference bus angles, voltage magnitudes (for AC) and generator injections.

6.1 Standard AC OPF

The optimization vector x for the standard AC OPF problem consists of the $n_b \times 1$ vectors of voltage angles Θ and magnitudes V_m and the $n_g \times 1$ vectors of generator real and reactive power injections P_g and Q_g .

$$x = \begin{bmatrix} \Theta \\ V_m \\ P_g \\ Q_g \end{bmatrix} \tag{6.5}$$

The objective function f(x) in (6.1) is simply a summation of individual polynomial cost functions f_P^i and f_Q^i of real and reactive power injections, respectively, for each generator:

$$f(P_g, Q_g) = \sum_{i=1}^{n_g} f_P^i(p_g^i) + f_Q^i(q_g^i).$$
(6.6)

The equality constraints in (6.2) are simply the full set of $2 \cdot n_b$ nonlinear real and reactive power balance equations from (4.2) and (4.3).

$$g_P(\Theta, V_m, P_g) = P_{\text{bus}}(\Theta, V_m) + P_d - C_g P_g = 0$$
(6.7)

$$g_Q(\Theta, V_m, Q_q) = Q_{\text{bus}}(\Theta, V_m) + Q_d - C_q Q_q = 0$$

$$(6.8)$$

The inequality constraints (6.3) consist of two sets of n_l branch flow limits as nonlinear functions of the bus voltage angles and magnitudes, one for the from end and one for the to end of each branch:

$$h_f(\Theta, V_m) = |F_f(\Theta, V_m)| - F_{\text{max}} \le 0 \tag{6.9}$$

$$h_t(\Theta, V_m) = |F_t(\Theta, V_m)| - F_{\text{max}} \le 0.$$

$$(6.10)$$

The flows are typically apparent power flows expressed in MVA, but can be real power flows (in MW) or currents, 31 yielding the following three possible forms for the flow constraints:

$$F_f(\Theta, V_m) = \begin{cases} S_f(\Theta, V_m), & \text{apparent power} \\ P_f(\Theta, V_m), & \text{real power} \\ I_f(\Theta, V_m), & \text{current} \end{cases}$$
(6.11)

where I_f is defined in (3.9), S_f in (3.15), $P_f = \Re\{S_f\}$ and the vector of flow limits F_{max} has the appropriate units for the type of constraint. It is likewise for $F_t(\Theta, V_m)$. The values used by MATPOWER'S OPF for the flow limits F_{max} are specified in the RATE_A column (6) of the branch matrix, 32 and the selection of flow constraint type in (6.11) is determined by the opf.flow_lim option.

The variable limits (6.4) include an equality constraint on any reference bus angle and upper and lower limits on all bus voltage magnitudes and real and reactive generator injections:

$$\theta_i^{\text{ref}} \le \theta_i \le \theta_i^{\text{ref}}, \qquad i \in \mathcal{I}_{\text{ref}}$$
 (6.12)

$$\theta_i^{\text{ref}} \leq \theta_i \leq \theta_i^{\text{ref}}, \qquad i \in \mathcal{I}_{\text{ref}}$$

$$v_m^{i,\min} \leq v_m^i \leq v_m^{i,\max}, \qquad i = 1 \dots n_b$$

$$p_g^{i,\min} \leq p_g^i \leq p_g^{i,\max}, \qquad i = 1 \dots n_g$$

$$q_g^{i,\min} \leq q_g^i \leq q_g^{i,\max}, \qquad i = 1 \dots n_g.$$

$$(6.12)$$

$$p_q^{i,\min} \le p_q^i \le p_q^{i,\max}, \qquad i = 1 \dots n_q \tag{6.14}$$

$$q_q^{i,\min} \le q_q^i \le q_q^{i,\max}, \qquad i = 1 \dots n_g. \tag{6.15}$$

The voltage reference angle θ_i^{ref} and voltage magnitude bounds $v_m^{i,\text{max}}$ and $v_m^{i,\text{min}}$ are specified in columns VA (9), VMAX (12) and VMIN (13), respectively, of row i of the bus matrix. Similarly, the generator bounds $q_g^{i,\text{max}}$, $q_g^{i,\text{min}}$, $p_g^{i,\text{max}}$ and $p_g^{i,\text{min}}$ are specified in columns QMAX (4), QMIN (5), PMAX (9) and PMIN (10), respectively, of row i of the gen matrix.

³¹For current constraints, the limit is expressed as an apparent power (MVA) value at a 1 p.u. voltage. This means the units of the specified limit and corresponding currents displayed in the output is in kiloamps $\times V_{\text{basekV}}$). To put it another way, the corresponding current in kiloamps can be found by dividing by the per unit voltage base in kV.

³²Setting the RATE_A column (6) of branch to zero is the preferred way to indicate a completely unconstrained line.

6.1.1Cartesian vs. Polar Coordinates for Voltage

Another variation of the standard AC OPF problem represents the bus voltages in cartesian, rather than polar, coordinates. That is, instead of Θ and V_m , the optimization vector x includes the real and imaginary parts of the complex voltage, denoted respectively by U and W, where V = U + jW.

$$x = \begin{bmatrix} U \\ W \\ P_g \\ Q_q \end{bmatrix} \tag{6.16}$$

The objective function remains unchanged, but the nodal power balance constraints (6.7) and (6.8) and branch flow constraints (6.9) and (6.10) are implemented as functions of U and W.

$$g_P(U, W, P_q) = P_{\text{bus}}(U, W) + P_d - C_q P_q = 0$$
 (6.17)

$$g_Q(U, W, Q_q) = Q_{\text{bus}}(U, W) + Q_d - C_q Q_q = 0$$
 (6.18)

$$h_f(U, W) = |F_f(U, W)| - F_{\text{max}} \le 0$$
 (6.19)

$$h_t(U, W) = |F_t(U, W)| - F_{\text{max}} \le 0.$$
 (6.20)

In this formulation, the voltage angle reference constraint (6.12) and voltage magnitude limits (6.13) cannot be simply applied as bounds on optimization variables. These constrained quantities also become functions of U and W.

$$\theta_i^{\text{ref}} \le \theta_i(u_i, w_i) \le \theta_i^{\text{ref}}, \qquad i \in \mathcal{I}_{\text{ref}}$$
 (6.21)

$$\theta_i^{\text{ref}} \le \theta_i(u_i, w_i) \le \theta_i^{\text{ref}}, \qquad i \in \mathcal{I}_{\text{ref}}$$

$$v_m^{i,\min} \le v_m^i(u_i, w_i) \le v_m^{i,\max}, \qquad i = 1 \dots n_b$$
(6.21)

In MATPOWER setting the opf.v_cartesian option to 1 (0 by default) selects the cartesian representation for voltages when running an AC OPF.³³

6.1.2 Current vs. Power for Nodal Balance Constraints

Another variation of the standard AC OPF problem uses current balance constraints in place of the power balance constraints (6.7)–(6.8) or (6.17)–(6.18). If we let M

 $^{^{33}}$ This option only applies to solvers based on MIPS, fmincon, IPOPT and Artelys Knitro.

and N represent the real and imaginary parts, respectively, of the current, we can express the current balance functions for the polar form as

$$g_M(\Theta, V_m, P_q, Q_q) = \Re\{I_{\text{bus}}(\Theta, V_m) + [V^*]^{-1}(S_d - C_q S_q)^*\} = 0$$
(6.23)

$$g_N(\Theta, V_m, P_q, Q_q) = \Im\{I_{\text{bus}}(\Theta, V_m) + [V^*]^{-1}(S_d - C_q S_q)^*\} = 0$$
 (6.24)

and for the cartesian form as

$$g_M(U, W, P_q, Q_q) = \Re\{I_{\text{bus}}(U, W) + [V^*]^{-1}(S_d - C_q S_q)^*\} = 0$$
 (6.25)

$$g_N(U, W, P_g, Q_g) = \Im\{I_{\text{bus}}(U, W) + [V^*]^{-1}(S_d - C_g S_g)^*\} = 0$$
 (6.26)

where $S_d = P_d + jQ_d$, $S_g = P_g + jQ_g$ and $[V^*]^{-1}$ is a diagonal matrix whose *i*-th diagonal entry is $1/v_i^*$, that is $\frac{1}{v_m^i}e^{j\theta_i}$ or $1/(u_i - jw_i)$.

In this formulation, which can be selected by setting the opf.current_balance option to 1,³⁴ the objective function and other constraints are not affected. This option can be used in conjunction with either the polar or cartesian representation of bus voltages.

6.2 Standard DC OPF

When using DC network modeling assumptions and limiting polynomial costs to second order, the standard OPF problem above can be simplified to a quadratic program, with linear constraints and a quadratic cost function. In this case, the voltage magnitudes and reactive powers are eliminated from the problem completely and real power flows are modeled as linear functions of the voltage angles. The optimization variable is

$$x = \begin{bmatrix} \Theta \\ P_a \end{bmatrix} \tag{6.27}$$

and the overall problem reduces to the following form.

$$\min_{\Theta, P_g} \sum_{i=1}^{n_g} f_P^i(p_g^i) \tag{6.28}$$

subject to

$$g_P(\Theta, P_g) = B_{\text{bus}}\Theta + P_{\text{bus,shift}} + P_d + G_{sh} - C_g P_g = 0$$
(6.29)

$$h_f(\Theta) = B_f \Theta + P_{f,\text{shift}} - F_{\text{max}} < 0 \tag{6.30}$$

$$h_t(\Theta) = -B_f\Theta - P_{f,\text{shift}} - F_{\text{max}} \le 0 \tag{6.31}$$

³⁴This option only applies to solvers based on MIPS, fmincon, IPOPT and Artelys Knitro.

$$\theta_i^{\text{ref}} \le \theta_i \le \theta_i^{\text{ref}}, \qquad i \in \mathcal{I}_{\text{ref}}$$
 (6.32)

$$\theta_i^{\text{ref}} \le \theta_i \le \theta_i^{\text{ref}}, \qquad i \in \mathcal{I}_{\text{ref}}$$

$$p_g^{i,\text{min}} \le p_g^i \le p_g^{i,\text{max}}, \qquad i = 1 \dots n_g$$
(6.32)

6.3 Extended OPF Formulation

MATPOWER employs an extensible OPF structure to allow the user to modify or augment the problem formulation without rewriting the portions that are shared with the standard OPF formulation described above. The standard formulation is modified by introducing additional variables, user-defined costs, and/or user-defined constraints. The full extended formulation can be written as follows:

$$\min_{\hat{x}} f(x) + f_u(\hat{x}) \tag{6.34}$$

subject to

$$\hat{g}(\hat{x}) = 0 \tag{6.35}$$

$$\hat{h}(\hat{x}) \le 0 \tag{6.36}$$

$$\hat{x}_{\min} \le \hat{x} \le \hat{x}_{\max} \tag{6.37}$$

$$l \le A\hat{x} \le u \tag{6.38}$$

The first difference to note is that the optimization variable x from the standard OPF formulation has been augmented with additional variables z to form a new optimization variable \hat{x} , and likewise with the lower and upper bounds.

$$\hat{x} = \begin{bmatrix} x \\ z \end{bmatrix} \quad \hat{x}_{\min} = \begin{bmatrix} x_{\min} \\ z_{\min} \end{bmatrix} \quad \hat{x}_{\max} = \begin{bmatrix} x_{\max} \\ z_{\max} \end{bmatrix}$$
 (6.39)

Second, there is an additional user-defined cost term $f_u(\hat{x})$ in the objective function. This cost consists of three pieces that will be described in more detail below.

$$f_u(\hat{x}) = f_q(\hat{x}) + f_{\text{nln}}(\hat{x}) + f_{\text{legacy}}(\hat{x})$$
 (6.40)

Third, the nonlinear constraints q and h are augmented with user defined additions g_u and h_u to give \hat{g} and \hat{h} .

$$\hat{g}(\hat{x}) = \begin{bmatrix} g(x) \\ g_u(\hat{x}) \end{bmatrix}, \quad \hat{h}(\hat{x}) = \begin{bmatrix} h(x) \\ h_u(\hat{x}) \end{bmatrix}$$
(6.41)

And finally, a new set of linear constraints are included in (6.38).

Up through version 6.0 of MATPOWER, the OPF extensions were handled via optional input parameters that define any additional variables,³⁵ linear constraints,³⁶ and costs of the pre-specified form defined by f_{legacy} .³⁷ This preserved the ability to use solvers that employ pre-compiled MEX files to compute all of the costs and constraints. This is referred to as MATPOWER's legacy extended OPF formulation [27].

For the AC OPF, subsequent versions also include the general nonlinear constraints g_u and h_u , and the quadratic and general nonlinear costs f_q and f_{nln} . The new quadratic cost terms can be handled by all of MATPOWER'S AC OPF solvers, but the general nonlinear costs and constraints require a solver that uses MATLAB code to implement the function, gradient and Hessian evaluations.³⁸

Section 7 describes the mechanisms available to the user for taking advantage of the extensible formulation described here.

6.3.1 User-defined Variables

The creation of additional user-defined z variables can be done explicitly or implicitly based on the difference between the number of columns in A and the dimension of the standard OPF optimization variable x. The optional vectors z_{\min} and z_{\max} are available to impose lower and upper bounds on z, respectively.

6.3.2 User-defined Constraints

- Linear Constraints The user-defined linear constraints (6.38) are general linear restrictions involving all of the optimization variables and are specified via matrix A and lower and upper bound vectors l and u. These parameters can be used to create equality constraints ($l_i = u_i$) or inequality constraints that are bounded below ($u_i = \infty$), bounded above ($l_i = \infty$) or bounded on both sides.
- Nonlinear Constraints The user-defined general nonlinear constraints take the form

$$g_u^j(x) = 0 \quad \forall j \in \mathcal{G}_u \tag{6.42}$$

$$h_u^j(x) \le 0 \quad \forall j \in \mathcal{H}_u,$$
 (6.43)

³⁵Parameters z_{\min} and z_{\max} in (6.39).

³⁶Parameters A, l, u in (6.38).

³⁷Parameters $H, C, N, \hat{r}, k, d, m$ in (6.48)–(6.51).

³⁸At the time of this writing, this includes solvers based on MIPS, fmincon, IPOPT and Artelys Knitro.

where \mathcal{G}_u and \mathcal{H}_u are sets of indices for user-defined equality and inequality constraint sets, respectively.

Each of these constraint sets is defined by two M-file functions, similar to those required by MIPS, one that computes the constraint values and their gradients (Jacobian), and the other that computes Hessian values.

6.3.3 User-defined Costs

The user-defined cost function f_u consists of three terms for three different types of costs: quadratic, general nonlinear, and legacy. Each term is a simple summation over all of the cost sets of that type.

$$f_{\mathbf{q}}(\hat{x}) = \sum_{i} f_{\mathbf{q}}^{j}(\hat{x}) \tag{6.44}$$

$$f_{\text{nln}}(\hat{x}) = \sum_{j} f_{\text{nln}}^{j}(\hat{x}) \tag{6.45}$$

$$f_{\text{legacy}}(\hat{x}) = \sum_{j} f_{\text{legacy}}^{j}(\hat{x})$$
 (6.46)

• Quadratic Costs for a cost set j are specified by parameters Q_j , c_j and k_j that define a quadratic function of the optimization variable \hat{x} .

$$f_{\mathbf{q}}^{j}(\hat{x}) = \hat{x}^{\mathsf{T}} Q_{j} \hat{x} + c_{j}^{\mathsf{T}} \hat{x} + k_{j} \tag{6.47}$$

- General Nonlinear Costs for a cost set j consist of a cost function $f_{\text{nln}}^{j}(\hat{x})$ provided in the form of a function handle to an M-file function that evaluates the cost and its gradients and Hessian for a given value of \hat{x} .
- Legacy Costs $f_{\text{legacy}}^{j}(\hat{x})$ are specified in terms of parameters H_{j} , C_{j} , N_{j} , \hat{r}_{j} , k_{j} , d_{j} and m_{j} . For simplicity of presentation, we will drop the j subscript for the rest of this discussion. All of the parameters are $n_{w} \times 1$ vectors except the symmetric $n_{w} \times n_{w}$ matrix H and the $n_{w} \times n_{\hat{x}}$ matrix N. The legacy user cost function takes the form

$$f_{\text{legacy}}(\hat{x}) = \frac{1}{2} w^{\mathsf{T}} H w + C^{\mathsf{T}} w \tag{6.48}$$

where w is defined in several steps as follows. First, a new vector u is created by applying a linear transformation N and shift \hat{r} to the full set of optimization variables

$$u = N\hat{x} - \hat{r},\tag{6.49}$$

then a scaled function with a "dead zone" is applied to each element of u to produce the corresponding element of w.

$$w_{i} = \begin{cases} m_{i} f_{d_{i}}(u_{i} + k_{i}), & u_{i} < -k_{i} \\ 0, & -k_{i} \leq u_{i} \leq k_{i} \\ m_{i} f_{d_{i}}(u_{i} - k_{i}), & u_{i} > k_{i} \end{cases}$$

$$(6.50)$$

Here k_i specifies the size of the "dead zone", m_i is a simple scale factor and f_{d_i} is a pre-defined scalar function selected by the value of d_i . Currently, MATPOWER implements only linear and quadratic options:

$$f_{d_i}(\alpha) = \begin{cases} \alpha, & \text{if } d_i = 1\\ \alpha^2, & \text{if } d_i = 2 \end{cases}$$
 (6.51)

as illustrated in Figure 6-1 and Figure 6-2, respectively.

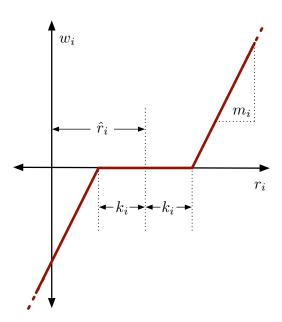


Figure 6-1: Relationship of w_i to r_i for $d_i = 1$ (linear option)

This form for f_{legacy} provides the flexibility to handle a wide range of costs, from simple linear functions of the optimization variables to scaled quadratic penalties on quantities, such as voltages, lying outside a desired range, to functions of linear combinations of variables, inspired by the requirements of price coordination terms found in the decomposition of large loosely coupled problems encountered in our own research.

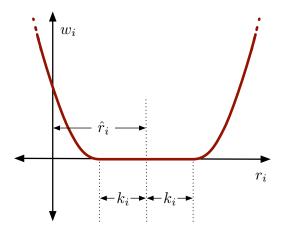


Figure 6-2: Relationship of w_i to r_i for $d_i = 2$ (quadratic option)

Some limitations are imposed on the parameters in the case of the DC OPF since MATPOWER uses a generic quadratic programming (QP) solver for the optimization. In particular, $k_i = 0$ and $d_i = 1$ for all i, so the "dead zone" is not considered and only the linear option is available for f_{d_i} . As a result, for the DC case (6.50) simplifies to $w_i = m_i u_i$.

6.4 Standard Extensions

In addition to making this extensible OPF structure available to end users, MAT-POWER also takes advantage of it internally to implement several additional capabilities.

6.4.1 Piecewise Linear Costs

The standard OPF formulation in (6.1)–(6.4) does not directly handle the non-smooth piecewise linear cost functions that typically arise from discrete bids and offers in electricity markets. When such cost functions are convex, however, they can be modeled using a constrained cost variable (CCV) method. The piecewise linear cost function c(x) is replaced by a helper variable y and a set of linear constraints that form a convex "basin" requiring the cost variable y to lie in the epigraph of the function c(x).

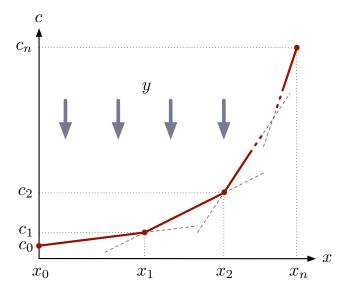


Figure 6-3: Constrained Cost Variable

Figure 6-3 illustrates a convex n-segment piecewise linear cost function

$$c(x) = \begin{cases} m_1(x - x_1) + c_1, & x \le x_1 \\ m_2(x - x_2) + c_2, & x_1 < x \le x_2 \\ \vdots & \vdots \\ m_n(x - x_n) + c_n, & x_{n-1} < x \end{cases}$$
(6.52)

defined by a sequence of points (x_j, c_j) , $j = 0 \dots n$, where m_j denotes the slope of the j-th segment

$$m_j = \frac{c_j - c_{j-1}}{x_j - x_{j-1}}, \quad j = 1 \dots n$$
 (6.53)

and $x_0 < x_1 < \cdots < x_n$ and $m_1 \le m_2 \le \cdots < m_n$.

The "basin" corresponding to this cost function is formed by the following n constraints on the helper cost variable y:

$$y \ge m_j(x - x_j) + c_j, \quad j = 1 \dots n.$$
 (6.54)

The cost term added to the objective function in place of c(x) is simply the variable y. MATPOWER uses this CCV approach internally to automatically generate the appropriate helper variable, cost term and corresponding set of constraints for any piecewise linear costs on real or reactive generation. All of MATPOWER'S OPF

solvers, for both AC and DC OPF problems, use the CCV approach with the exception of two that are part of the optional TSPOPF package [28], namely the step-controlled primal/dual interior point method (SCPDIPM) and the trust region based augmented Lagrangian method (TRALM), both of which use a cost smoothing technique instead [8].

Note that MATPOWER (in opf_setup) automatically converts any single-segment piecewise linear costs into polynomial (linear) form to avoid unnecessarily creating extra variables and constraints. It is this modified cost, rather than the original piecewise linear equivalent, that is returned in the gencost field of the results struct.

6.4.2 Dispatchable Loads

A simple approach to dispatchable or price-sensitive loads is to model them as negative real power injections with associated negative costs. This is done by specifying a generator with a negative output, ranging from a minimum injection equal to the negative of the largest possible load to a maximum injection of zero.

Consider the example of a price-sensitive load whose marginal benefit function is shown in Figure 6-4. The demand p_d of this load will be zero for prices above λ_1 , p_1 for prices between λ_1 and λ_2 , and $p_1 + p_2$ for prices below λ_2 .

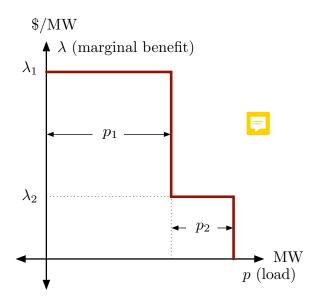


Figure 6-4: Marginal Benefit or Bid Function

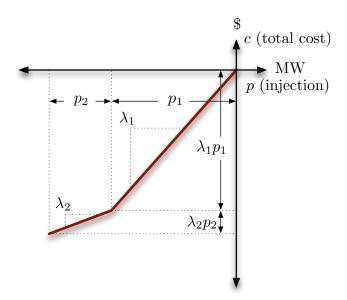


Figure 6-5: Total Cost Function for Negative Injection

This corresponds to a negative generator with the piecewise linear cost curve shown in Figure 6-5. Note that this approach assumes that the demand blocks can be partially dispatched or "split". Requiring blocks to be accepted or rejected in their entirety would pose a mixed-integer problem that is beyond the scope of the current MATPOWER implementation.

It should be noted that, with this definition of dispatchable loads as negative generators, if the negative cost corresponds to a benefit for consumption, minimizing the cost f(x) of generation is equivalent to maximizing social welfare.

With an AC network model, there is also the question of reactive dispatch for such loads. Typically the reactive injection for a generator is allowed to take on any value within its defined limits. Since this is not normal load behavior, the model used in MATPOWER assumes that dispatchable loads maintain a constant power factor. When formulating the AC OPF problem, MATPOWER will automatically generate an additional equality constraint to enforce a constant power factor for any "negative generator" being used to model a dispatchable load.

The power factor, which can be lagging or leading, is determined by the ratio of reactive to active power for the load and is specified by the active and reactive limits defining the nominal load in the gen matrix. For all dispatchable loads, the value in the PMIN column is negative and defines the nominal active power of the load. And PMAX is zero, allowing the load to be fully curtailed depending on the price.

The reactive limits, QMIN and QMAX, depend on whether the power flow is lagging or leading. One defines the nominal reactive load and the other must be zero to allow the load to be fully curtailed. The values of PG and QG must be defined to be consistent with the nominal power factor.

- Lagging Power Factor The reactive injection is negative, meaning that reactive power is consumed by the load. Hence, QMIN is negative, QMAX is zero, and PG and QG must be set so that QG is equal to PG * QMIN/PMIN.
- Leading Power Factor The reactive injection is positive, that is, reactive power is produced by the load. Hence, QMAX is positive, QMIN is zero, and PG and QG must be set so that QG is equal to PG * QMAX/PMIN.

6.4.3 Generator Capability Curves

The typical AC OPF formulation includes box constraints on a generator's real and reactive injections, specified as simple lower and upper bounds on p (p_{\min} and p_{\max}) and q (q_{\min} and q_{\max}). On the other hand, the true P-Q capability curves of physical generators usually involve some tradeoff between real and reactive capability, so that it is not possible to produce the maximum real output and the maximum (or minimum) reactive output simultaneously. To approximate this tradeoff, MAT-POWER includes the ability to add an upper and lower sloped portion to the standard box constraints as illustrated in Figure 6-6, where the shaded portion represents the feasible operating region for the unit.

The two sloped portions are constructed from the lines passing through the two pairs of points defined by the six parameters p_1 , q_1^{\min} , q_1^{\max} , p_2 , q_2^{\min} , and q_2^{\max} . If these six parameters are specified for a given generator in columns PC1-QC2MAX (11-16), MATPOWER automatically constructs the corresponding additional linear inequality constraints on p and q for that unit.

If one of the sloped portions of the capability constraints is binding for generator k, the corresponding shadow price is decomposed into the corresponding $\mu_{P_{\max}}$ and $\mu_{Q_{\min}}$ or $\mu_{Q_{\max}}$ components and added to the respective column (MU_PMAX, MU_QMIN or MU_QMAX) in the k^{th} row of gen.

6.4.4 Branch Angle Difference Limits

The difference between the bus voltage angle θ_f at the from end of a branch and the angle θ_t at the to end can be bounded above and below to act as a proxy for a transient stability limit, for example. If these limits are provided in columns ANGMIN

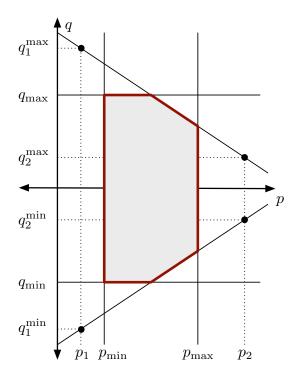


Figure 6-6: Generator P-Q Capability Curve

(12) and ANGMAX (13) of the branch matrix, MATPOWER creates the corresponding constraints on the voltage angle variables.³⁹

6.5 Solvers

Early versions of MATPOWER relied on MATLAB's Optimization Toolbox [29] to provide the NLP and QP solvers needed to solve the AC and DC OPF problems, respectively. While they worked reasonably well for very small systems, they did not scale well to larger networks. Eventually, optional packages with additional solvers were added to improve performance, typically relying on MATLAB extension (MEX) files implemented in Fortran or C and pre-compiled for each machine architecture. Some of these MEX files are distributed as optional packages due to differences in terms of use. For DC optimal power flow, there is a MEX build [30] of the high

³⁹The voltage angle difference for branch k is taken to be unbounded below if branch(k, ANGMIN) is less than or equal to -360 and unbounded above if branch(k, ANGMAX) is greater than or equal to 360. If both parameters are zero, the voltage angle difference is unconstrained.

performance interior point BPMPD solver [31] for LP/QP problems. For the AC OPF problem, the MINOPF [32] and TSPOPF [28] packages provide solvers suitable for much larger systems. The former is based on MINOS [33] and the latter includes the primal-dual interior point and trust region based augmented Lagrangian methods described in [8]. MATPOWER version 4 and later also includes the option to use the open-source IPOPT solver⁴⁰ for solving both AC and DC OPFs, based on the MATLAB MEX interface to IPOPT⁴¹. It also includes the option to use CPLEX⁴² or MOSEK⁴³ for DC OPFs. MATPOWER 4.1 added the option to use Artelys Knitro [34]⁴⁴ for AC OPFs and the Gurobi Optimizer [35]⁴⁵ for DC OPFs and MATPOWER 5 added GLPK [36] and 5.1 added CLP [37]. See Appendix G for more details on these optional packages.

Beginnning with version 4, MATPOWER also includes its own primal-dual interior point method implemented in pure-MATLAB code, derived from the MEX implementation of the algorithms described in [8]. This solver is called MIPS (MATPOWER Interior Point Solver) and is described in more detail in Appendix A. If no optional packages are installed, MIPS will be used by default for both the AC OPF and as the QP solver used by the DC OPF. The AC OPF solver also employs a unique technique for efficiently forming the required Hessians via a few simple matrix operations [38–40]. This solver has application to general nonlinear optimization problems outside of MATPOWER and can be called directly as mips. There is also a convenience wrapper function called qps_mips making it trivial to set up and solve LP and QP problems, with an interface similar to quadprog from the MATLAB Optimization Toolbox.

6.6 runopf

In MATPOWER, an optimal power flow is executed by calling runopf with a case struct or case file name as the first argument (casedata). In addition to printing output to the screen, which it does by default, runopf optionally returns the solution in a results struct.

```
>> results = runopf(casedata);
```

⁴⁰Available from https://github.com/coin-or/Ipopt.

 $^{^{41}} See\ https://projects.coin-or.org/Ipopt/wiki/MatlabInterface.$

⁴²See https://www.ibm.com/analytics/cplex-optimizer.

⁴³See https://www.mosek.com/.

⁴⁴See https://www.artelys.com/solvers/knitro/.

⁴⁵See https://www.gurobi.com/.

The results struct is a superset of the input Matpower case struct mpc, with some additional fields as well as additional columns in some of the existing data fields. In addition to the solution values included in the results for a simple power flow, shown in Table 4-1 in Section 4.4, the following additional optimal power flow solution values are stored as shown in Table 6-1.

Table 6-1: Optimal Power Flow Results

name	description
results.f	final objective function value
results.x	final value of optimization variables (internal order)
results.om	OPF model object [†]
results.bus(:, LAM_P)	Lagrange multiplier on real power mismatch
results.bus(:, LAM_Q)	Lagrange multiplier on reactive power mismatch
results.bus(:, MU_VMAX)	Kuhn-Tucker multiplier on upper voltage limit
results.bus(:, MU_VMIN)	Kuhn-Tucker multiplier on lower voltage limit
results.gen(:, MU_PMAX)	Kuhn-Tucker multiplier on upper P_g limit
results.gen(:, MU_PMIN)	Kuhn-Tucker multiplier on lower P_g limit
results.gen(:, MU_QMAX)	Kuhn-Tucker multiplier on upper Q_g limit
results.gen(:, MU_QMIN)	Kuhn-Tucker multiplier on lower Q_g limit
results.branch(:, MU_SF)	Kuhn-Tucker multiplier on flow limit at "from" bus
results.branch(:, MU_ST)	Kuhn-Tucker multiplier on flow limit at "to" bus
results.mu	shadow prices of constraints [‡]
results.g	(optional) constraint values
results.dg	(optional) constraint 1st derivatives
results.raw	raw solver output in form returned by MINOS, and more [‡]
results.var.val	final value of optimization variables, by named subset [‡]
results.var.mu	shadow prices on variable bounds, by named subset [‡]
results.nle	shadow prices on nonlinear equality constraints, by named subset [‡]
results.nli	shadow prices on nonlinear inequality constraints, by named subset [‡]
results.lin	shadow prices on linear constraints, by named subset [‡]
results.cost	final value of user-defined costs, by named subset [‡]

 $^{^\}dagger$ See help for opf_model and opt_model for more details. ‡ See help for opf for more details.

Additional optional input arguments can be used to set options (mpopt) and provide file names for saving the pretty printed output (fname) or the solved case data (solvedcase).

```
>> results = runopf(casedata, mpopt, fname, solvedcase);
```

Some of the main options that control the optimal power flow simulation are listed in Tables 6-2 and 6-3. There are many other options that can be used to control the termination criteria and other behavior of the individual solvers. See Appendix C or the mpoption help for details. As with runpf the output printed to the screen can be controlled by the options in Table 4-3, but there are additional output options for the OPF, related to the display of binding constraints that are listed Table 6-4, along with an option that can be used to force the AC OPF to return information about the constraint values and Jacobian and the objective function gradient and Hessian.

For OPF problems, the preferred way to eliminate the flow limit on a branch is to set the RATE_A column (6) of the corresponding row in the branch matrix to zero. This indicates a completely unconstrained flow (as opposed to zero flow).

By default, runopf solves an AC optimal power flow problem using a primal dual interior point method. To run a DC OPF, the model option must be set to 'DC'. For convenience, MATPOWER provides a function rundcopf which is simply a wrapper that sets the model to 'DC' before calling runopf.

Internally, the runopf function does a number of conversions to the problem data before calling the appropriate solver routine for the selected OPF algorithm. This external-to-internal format conversion is performed by the ext2int function, described in more detail in Section 7.3.1, and includes the elimination of out-of-service equipment and the consecutive renumbering of buses. All computations are done using this internal indexing. When the simulation has completed, the data is converted back to external format by int2ext before the results are printed and returned. In addition, both ext2int and int2ext can be customized via user-supplied callback routines to convert data needed by user-supplied variables, constraints or costs into internal indexing.

Table 6-2: Optimal Power Flow Solver Options

opf.ac.solver 'DEF	FAULT'	AC optimal power flow solver:
		IDECAMEN -1 1-f1+1 : - IMIDC'
		'DEFAULT' – choose default solver, i.e. 'MIPS'
		'MIPS' - MIPS, MATPOWER Interior Point Solver,
		primal/dual interior point method [†]
		'FMINCON' - MATLAB Optimization Toolbox, fmincon
		'IPOPT' - IPOPT*
		'KNITRO' - Artelys Knitro*
		'MINOPF' - MINOPF, MINOS-based solver
		'PDIPM' – PDIPM, primal/dual interior point method [‡]
		'SDPOPF' - SDPOPF, solver based on semidefinite relaxation
		'TRALM' - TRALM,* trust region based augmented Lan-
		grangian method
opf.dc.solver 'DEF	' AULT	DC optimal power flow solver:
		'DEFAULT' - choose default solver based on availability
		in the following order: 'GUROBI', 'CPLEX',
		'MOSEK', 'OT', 'GLPK' (linear costs only),
		'BPMPD', 'MIPS'
		'MIPS' - MIPS, MATPOWER Interior Point Solver,
		primal/dual interior point method †
		'BPMPD' - BPMPD*
		'CLP' - CLP*
		'CPLEX' - CPLEX*
		$\verb"GLPK" - GLPK" (no \ quadratic \ costs)$
		'GUROBI' - Gurobi*
		'IPOPT' - IPOPT*
		'MOSEK' - MOSEK*
		'OT' — MATLAB Opt Toolbox, quadprog, linprog

^{*} Requires the installation of an optional package. See Appendix G for details on the corresponding package.
† For MIPS-sc, the step-controlled version of this solver, the mips.step_control option must be set to 1.
‡ For SC-PDIPM, the step-controlled version of this solver, the pdipm.step_control option must be set to 1.

Table 6-3: Other OPF Options

name	default	description
opf.current_balance [‡]	0	use current, as opposed to power, balance formulation for AC OPF, 0 or 1
$\mathtt{opf.v_cartesian}^\ddagger$	0	use cartesian, as opposed to polar, representation for voltages for AC OPF, 0 or 1
opf.violation	5×10^{-6}	constraint violation tolerance
opf.use_vg	0	respect generator voltage setpoint, 0 or 1 [†] 0 – use voltage magnitude limits specified in bus, ignore VG in gen 1 – replace voltage magnitude limits specified in bus by VG in corresponding gen
opf.flow_lim	'S'	quantity to limit for branch flow constraints 'S' - apparent power flow (limit in MVA) 'P' - active power flow (limit in MW) 'I' - current magnitude (limit in MVA at 1 p.u. voltage, i.e. kA·V _{basekV}) '2' - same as 'P', but implemented using square of active flow, rather than simple max
opf.ignore_angle_lim	0	ignore angle difference limits for branches 0 – include angle difference limits, if specified 1 – ignore angle difference limits even if specified
opf.softlims.default	1	behavior of OPF soft limits for which parameters are not explicitly provided 0 - do not include softlims if not explicitly specified 1 - include softlims with default values if not explicitly specified
opf.init_from_mpc*	-1	specify whether to use the current state in Matpower case to initialize OPF -1 - Matpower decides based on solver/algorithm 0 - ignore current state in Matpower case [‡] 1 - use current state in Matpower case
opf.start	0	strategy for initializing OPF starting point 0 – default, MATPOWER decides based on solver, (currently identical to 1) 1 – ignore current state in MATPOWER case [‡] 2 – use current state in MATPOWER case 3 – solve power flow and use resulting state
opf.return_raw_der	0	for AC OPF, return constraint and derivative info in results.raw (in fields g, dg, df, d2f)

^{*} Deprecated. Use opf.start instead.
† Using a value between 0 and 1 results in the limits being determined by the corresponding weighted average of

[†] Only applies to fmincon, IPOPT, Artelys Knitro and MIPS solvers, which use an interior point estimate; others use current state in Matpower case, as with opf.start = 2.

Table 6-4: OPF Output Options

name	default	description	
out.lim.all	-1	controls constraint info output -1 – individual flags control what is printed 0 – do <i>not</i> print any constraint info [†] 1 – print only binding constraint info [†] 2 – print all constraint info [†]	
out.lim.v	1	control output of voltage limit info 0 – do not print 1 – print binding constraints only 2 – print all constraints	
out.lim.line	1	control output of line flow limit info [‡]	
out.lim.pg	1	control output of gen active power limit info [‡]	
out.lim.qg	1	control output of gen reactive power limit info [‡]	

[†] Overrides individual flags. † Takes values of 0, 1 or 2 as for out.lim.v.

7 Extending the OPF

The extended OPF formulation described in Section 6.3 allows the user to modify the standard OPF formulation to include additional variables, costs and/or constraints. There are two primary mechanisms available for the user to accomplish this. The first is by directly constructing the full parameters for the additional costs or constraints and supplying them either as fields in the case struct or directly as arguments to the opf function. The second, and more powerful, method is via a set of callback functions that customize the OPF at various stages of the execution. Matpower includes several examples of using the latter method, for example to add a fixed zonal reserve requirement, to implement interface flow limits, dispatchable DC transmission lines, or branch flow soft limits.

7.1 Direct Specification

This section describes the additional fields that can be added to mpc (the MATPOWER case struct) or, alternatively, the input parameters that can be passed directly to the opf function. The former is the preferred approach, as it allows the normal use of runopf and related functions.

7.1.1 User-defined Variables

In the case of direct specification, additional variables are created implicitly based on the difference between the number of columns in A and the number n_x of standard OPF variables. If A has more columns than x has elements, the extra columns are assumed to correspond to a new z variable. The initial value and lower and upper bounds for z can also be specified in the optional fields or arguments, z0, z1 and zu, respectively.

7.1.2 User-defined Constraints

- Linear Constraints can be added by specifying the A, l and u parameters of (6.38) directly as fields or arguments of the same names, A, 1 and u, respectively, where A is sparse.
- Nonlinear Constraints for a constraint set j, that is, $g_u^j(x)$ or $h_u^j(x)$ from (6.42) or (6.43), are implemented by defining two M-file functions, similar to

those required by MIPS. The first is a function to compute the constraint values and their gradients (Jacobian), 46 with calling syntax:

```
[g, dg] = my_constraint_fcn(x, <p1>, <p2>, ...)
```

Here $\langle p1 \rangle$ and $\langle p2 \rangle$ represent arbitrary optional input parameters that remain constant for all calls to the function. The second is a function to compute the Hessian of the corresponding term in the Lagrangian function, that is the term corresponding to $\lambda^{\mathsf{T}} g_u^j(x)$ or $\mu^{\mathsf{T}} h_u^j(x)$. This function has the following calling syntax:

```
d2G = my_constraint_hess(x, lambda, <p1>, <p2>, ...)
```

Once again this is similar to the form of the Hessian evaluation function expected by MIPS. 47

In order to add a set of user-defined nonlinear constraints to the formulation through direct specification, an entry must be added to the optional user_constraints.nle (nonlinear equality) or user_constraints.nli (nonlinear inequality) fields of the MATPOWER case struct. These fields are cell arrays, in which each element (also a cell array) specifies a constraint set of the corresponding kind. The format is as follows, where the details are described in Table 7-1.

```
mpc.user_constraints.nle = {
          {name, N, g_fcn, hess_fcn, varsets, params},
          ...
};
```

There is an example in the OPF testing code (e.g. t_opf_mips) near the end, that implements a nonlinear relationship between three different generator outputs by adding the following line.

⁴⁶The differences between this function and the one required by MIPS is that (1) this one represents a single set of equality constraints or inequality constraints, not both, (2) it returns in dg the $m \times n$ Jacobian (for m constraints and n variables), whereas MIPS expects the $n \times m$ transpose, (3) this function allows arbitrary additional parameters, and (4) x can be a cell array of sub-vectors of optimization variable \hat{x} (see varsets in Table 7-1).

⁴⁷The differences between this function and the one required by MIPS is that (1) this one returns the Hessian of a single term of the Lagrangian, not of the full Lagrangian, (2) the lambda argument is a simple vector of multipliers (λ or μ) corresponding just to this set of constraints, not a struct containing the full λ and μ vectors, in fields eqnonlin and ineqnonlin, respectively, and (3) this function allows arbitrary additional parameters, and (4) x can be a cell array of sub-vectors of optimization variable \hat{x} (see varsets in Table 7-1).

Table 7-1: User-defined Nonlinear Constraint Specification

name	description
name N g_fcn	string with name of constraint set, used to label multipliers in results struct number of constraints, i.e. dimension of $g_u^j(x)$ (or $h_u^j(x)$ as the case may be) string containing name of function to evaluate constraint and gradients (Jacobian)
hess_fcn	string containing name of function to evaluate Hessian of corresponding term of Lagrangian
varsets	cell array of variable set names, specifying sub-vectors of optimization vector \hat{x} to be passed as inputs in \mathbf{x}^{\dagger}
params	cell array of optional, arbitrary parameters to pass to each call to the constraint and Hessian evaluation functions

[†] If varsets is empty, x will be the full optimization vector \hat{x} , otherwise it will be a cell array of sub-vectors of \hat{x} for the specified variable sets. Valid names include 'Va', 'Vm', 'Pg', and 'Qg'. It can include others depending on the OPF extensions in use. See the variable names displayed by results.om for a complete list for your problem.

This adds a single constraint named 'Pg_usr' as a function of the vector of generator active injections (x will be $\{Pg\}$, where Pg is a sub-vector of the optimization vector \hat{x} containing only the generator active injections). The constraints and gradients are evaluated by a function named 'opf_nle_fcn1' and the Hessian of the corresponding term of the Lagrangian by a function named 'opf_nle_hess1', neither of which expect any additional parameters.

7.1.3 User-defined Costs

- Quadratic Costs by direct specification are not currently supported.
- General Nonlinear Costs by direct specification are not currently supported.
- Legacy Costs To add legacy costs directly, the parameters H, C, N, \hat{r}, k, d and m of (6.48)–(6.51) described in Section 6.3.3 are specified as fields or arguments H, Cw, N and fparm, respectively, where fparm is the $n_w \times 4$ matrix

$$f_{\text{parm}} = \left[\begin{array}{ccc} d & \hat{r} & k & m \end{array} \right]. \tag{7.1}$$

When specifying additional costs, N and Cw are required, while H and fparm are optional. The default value for H is a zero matrix, and the default for f_{parm} is

such that d and m are all ones and \hat{r} and k are all zeros, resulting in simple linear cost, with no shift or "dead-zone". N and H should be specified as sparse matrices.

7.1.4 Additional Comments

For a simple formulation extension to be used for a small number of OPF cases, this method has the advantage of being direct and straightforward. While MATPOWER does include code to eliminate the columns of A and N corresponding to V_m and Q_g when running a DC OPF⁴⁸, as well as code to reorder and eliminate columns appropriately when converting from external to internal data formats, this mechanism still requires the user to take special care in preparing the A and N matrices to ensure that the columns match the ordering of the elements of the optimization vectors x and z. All extra constraints and variables must be incorporated into a single set of parameters that are constructed before calling the OPF. The bookkeeping needed to access the resulting variables and shadow prices on constraints and variable bounds must be handled manually by the user outside of the OPF, along with any processing of additional input data and processing, printing or saving of the additional result data. Making further modifications to a formulation that already includes user-supplied costs, constraints or variables, requires that both sets be incorporated into a new single consistent set of parameters.

7.2 Callback Functions

The second method, based on defining a set of callback functions, offers several distinct advantages, especially for more complex scenarios or for adding a feature for others to use, such as the zonal reserve requirement or the interface flow limits mentioned previously. This approach makes it possible to:

- define and access variable/constraint/cost sets as individual named blocks
- define constraints, costs only in terms of variables directly involved
- pre-process input data and/or post-process result data
- print and save new result data
- simultaneously use multiple, independently developed extensions (e.g. zonal reserve requirements and interface flow limits)

⁴⁸Only if they contain all zeros.

With this approach the OPF formulation is modified in the formulation callback, which is described and illustrated below in Section 7.3.2. The modifications to the formulation are handled by adding variables, costs and/or constraints to the OPF Model object (om) using one of the add_* methods. Please see the documentation for each of these methods for more details.

7.2.1 User-defined Variables

Additional variables are added using the add_var method.

```
om.add_var('myV', N, myVO, myV_min, myV_max);
```

Here myV is the name of the variable set, N the number of variables being added (i.e. this is an $N \times 1$ vector being appended to the current optimization variable \hat{x}), and myVO, myV_min, myV_max are vectors of initial values, lower bounds and upper bounds, respectively.

7.2.2 User-defined Costs

• Quadratic Costs are added using the add_quad_cost method.

```
om.add_quad_cost('myQcost', Q, c, k, varsets);
```

Here myQcost is the name of the cost set, Q, c and k are the Q_j , c_j and k_j parameters from (6.47), respectively, and var_sets is an optional cell array of variable set names. If var_sets is provided, the dimensions of Q and c will correspond, not to the full optimization vector \hat{x} , but to the subvector formed by stacking only the variable sets specified in var_sets.

• General Nonlinear Costs are added using the add_nln_cost method.

```
om.add_nln_cost('myNonlinCost', 1, fcn, varsets);
```

Here myNonlinCost is the name of the cost set, fcn is a function handle and var_sets is an optional cell array of variable set names. The function handle fcn must point to a function that implements the $f_{\text{nln}}^{j}(\hat{x})$ function from (6.46), returning the value of the function, along with its gradient and Hessian. This function has a format similar to that required by MIPS for its cost function.

If var_sets is provided, instead of the full optimization vector \hat{x} , the x passed to the function fcn will be a cell array of sub-vectors of \hat{x} corresponding to the specified variable sets.

• Legacy Costs are added using the add_legacy_cost method.

```
om.add_legacy_cost('myLegacyCost', cp, varsets);
```

Here myLegacyCost is the name of the cost set, cp is a struct containing the cost parameters H, C, N, \hat{r} , k, d and m from (6.48)–(6.51) in fields H, Cw, N, rh, kk, dd and mm, respectively, and var_sets is an optional cell array of variable set names. If var_sets is provided, the number of columns in N will correspond, not to the full optimization vector \hat{x} , but to the subvector formed by stacking only the variable sets specified in var_sets.

7.2.3 User-defined Constraints

• Linear Constraints are added using the add_lin_constraint method.

```
om.add_lin_constraint('myLinCons', A, 1, u, varsets);
```

Here myLinCons is the name of the constraint set, A, 1 and u correspond to additional rows to add to the A, l and u parameters in (6.38), respectively, and var_sets is an optional cell array of variable set names. If var_sets is provided, the number of columns in A will correspond, not to the full optimization vector \hat{x} , but to the subvector formed by stacking only the variable sets specified in var_sets.

Nonlinear Constraints are added using the add_nln_constraint method.

```
om.add_nln_constraint('myNonlinCons', N, iseq, fcn, hess, varsets);
```

Here myNonlinCons is the name of the constraint set, N is the number of constraints being added, iseq is a true or false value indicating whether this is a set of equality (or else inequality) constraints, fcn and hess are function handles and var_sets is an optional cell array of variable set names. The function handle fcn must point to a function that implements the corresponding constraint function $g_u^j(x)$ or $h_u^j(x)$ and its gradients, and hess to one that evaluates the corresponding term in the Hessian of the Lagrangian function. This function has a format similar to that required by MIPS for its cost function. See more details in Section 7.1.2.

If var_sets is provided, instead of the full optimization vector \hat{x} , the x passed to the function fcn will be a cell array of sub-vectors of \hat{x} corresponding to the specified variable sets.

7.3 Callback Stages and Example

MATPOWER defines five stages in the execution of a simulation where custom code can be inserted to alter the behavior or data before proceeding to the next stage. This custom code is defined as a set of "callback" functions that are registered via add_userfcn for MATPOWER to call automatically at one of the five stages. Each stage has a name and, by convention, the name of a user-defined callback function ends with the name of the corresponding stage. For example, a callback for the formulation stage that modifies the OPF problem formulation to add reserve requirements could be registered with the following line of code.

mpc = add_userfcn(mpc, 'formulation', @userfcn_reserves_formulation);

The sections below will describe each stage and the input and output arguments for the corresponding callback function, which vary depending on the stage. An example that employs additional variables, constraints and costs will be used for illustration.

Consider the problem of jointly optimizing the allocation of both energy and reserves, where the reserve requirements are defined as a set of n_{rz} fixed zonal MW quantities. Let Z_k be the set of generators in zone k and R_k be the MW reserve requirement for zone k. A new set of variables r are introduced representing the reserves provided by each generator. The value r_i , for generator i, must be nonnegative and is limited above by a user-provided upper bound r_i^{max} (e.g. a reserve offer quantity) as well as the physical ramp rate Δ_i .

$$0 \le r_i \le \min(r_i^{\max}, \Delta_i), \quad i = 1 \dots n_g \tag{7.2}$$

If the vector c contains the marginal cost of reserves for each generator, the user defined cost term from (6.34) is simply

$$f_u(\hat{x}) = c^{\mathsf{T}} r. \tag{7.3}$$

There are two additional sets of constraints needed. The first ensures that, for each generator, the total amount of energy plus reserve provided does not exceed the capacity of the unit.

$$p_g^i + r_i \le p_g^{i,\text{max}}, \quad i = 1 \dots n_g \tag{7.4}$$

The second requires that the sum of the reserve allocated within each zone k meets the stated requirements.

$$\sum_{i \in Z_k} r_i \ge R_k, \quad k = 1 \dots n_{rz} \tag{7.5}$$

Table 7-2: Names Used by Implementation of OPF with Reserves

name	description
mpc	Matpower case struct
reserves	additional field in mpc containing input parameters for zonal reserves in
	the following sub-fields:
cost	$n_q \times 1$ vector of reserve costs, c from (7.3)
qty	$n_g \times 1$ vector of reserve quantity upper bounds, i^{th} element is r_i^{max}
zones	$n_{rz} \times n_q$ matrix of reserve zone definitions
	$\mathtt{zones(k,j)} = \left\{ egin{array}{ll} 1 & ext{if gen } j ext{ belongs to reserve zone } k & (j \in Z_k) \\ 0 & ext{otherwise} & (j \notin Z_k) \end{array} ight.$
req	$n_{rz} \times 1$ vector of zonal reserve requirements, k^{th} element is R_k from (7.5)
om	OPF model object, already includes standard OPF setup
results	OPF results struct, superset of mpc with additional fields for output data
ng	n_g , number of generators
R	name for new reserve variable block, i^{th} element is r_i
Pg_plus_R	name for new capacity limit constraint set (7.4)
Rreq	name for new reserve requirement constraint set (7.5)

Table 7-2 describes some of the variables and names that are used in the example callback function listings in the sections below.

7.3.1 ext2int Callback

Before doing any simulation of a case, MATPOWER performs some data conversion on the case struct in order to achieve a consistent internal structure, by calling the following.

```
mpc = ext2int(mpc, mpopt);
```

All isolated buses, out-of-service generators and branches are removed, along with any generators or branches connected to isolated buses and the buses are renumbered consecutively, beginning at 1. All of the related indexing information and the original data matrices are stored in an order field in the case struct to be used later by int2ext to perform the reverse conversions when the simulation is complete.

The first stage callback is invoked from within the ext2int function immediately after the case data has been converted. Inputs are a MATPOWER case struct (mpc) freshly converted to internal indexing, a MATPOWER options struct mpopt, 49 and any (optional) args value supplied when the callback was registered via add_userfcn.

 $^{^{49}}$ The mpopt may be empty if ext2int() is called manually without providing it.

Output is the (presumably updated) mpc. This is typically used to reorder any input arguments that may be needed in internal ordering by the formulation stage. The example shows how e2i_field can also be used, with a case struct that has already been converted to internal indexing, to convert other data structures by passing in 2 or 3 extra parameters in addition to the case struct. In this case, it automatically converts the input data in the qty, cost and zones fields of mpc.reserves to be consistent with the internal generator ordering, where off-line generators have been eliminated. Notice that it is the second dimension (columns) of mpc.reserves.zones that is being re-ordered. See the on-line help for e2i_field and e2i_data for more details on what all they can do.

```
function mpc = userfcn_reserves_ext2int(mpc, mpopt, args)

mpc = e2i_field(mpc, {'reserves', 'qty'}, 'gen');
mpc = e2i_field(mpc, {'reserves', 'cost'}, 'gen');
mpc = e2i_field(mpc, {'reserves', 'zones'}, 'gen', 2);
```

This stage is also a good place to check the consistency of any additional input data required by the extension and throw an error if something is missing or not as expected.

7.3.2 formulation Callback

This stage is called at the end of opf_setup after the OPF Model (om) object has been initialized with the standard OPF formulation, but before calling the solver. This is the ideal place for modifying the problem formulation with additional variables, constraints and costs, using the add_var, add_lin_constraint, add_nln_constraint, add_quad_cost, add_nln_cost and add_legacy_cost methods of the OPF Model object. Inputs are the om object, the MATPOWER options struct mpopt and any (optional) args supplied when the callback was registered via add_userfcn. Output is the updated om object.

The om object contains both the original MATPOWER case data as well as all of the indexing data for the variables and constraints of the standard OPF formulation. See the on-line help for opf_model and opt_model for more details on the OPF model object and the methods available for manipulating and accessing it.

⁵⁰It is perfectly legitimate to register more than one callback per stage, such as when enabling multiple independent OPF extensions. In this case, the callbacks are executed in the order they were registered with add_userfcn. E.g. when the second and subsequent formulation callbacks are invoked, the om object will reflect any modifications performed by earlier formulation callbacks.

In the example code, a new variable block named R with n_g elements and the limits from (7.2) is added to the model via the add_var method. Similarly, two linear constraint blocks named Pg_plus_R and Rreq, implementing (7.4) and (7.5), respectively, are added via the add_lin_constraint method. And finally, the add_quad_cost method is used to add to the model a quadratic (actually linear) cost block corresponding to (7.3).

Notice that the last argument to add_lin_constraint and add_quad_cost allows the constraints and costs to be defined only in terms of the relevant parts of the optimization variable \hat{x} . For example, the A matrix for the Pg_plus_R constraint contains only columns corresponding to real power generation (Pg) and reserves (R) and need not bother with voltages, reactive power injections, etc. As illustrated in Figure 7-1, this allows the same code to be used with both the AC OPF, where \hat{x} includes V_m and Q_g , and the DC OPF where it does not. This code is also independent of any additional variables that may have been added by MATPOWER (e.g. y variables from MATPOWER's CCV handling of piece-wise linear costs) or by the user via previous formulation callbacks. MATPOWER will place the constraint and cost matrix blocks in the appropriate place when it constructs the aggregated constraint and cost matrices at run-time. This is an important feature that enables independently developed MATPOWER OPF extensions to work together.

```
function om = userfcn_reserves_formulation(om, mpopt, args)
%% initialize some things
define_constants;
mpc = om.get_mpc();
r = mpc.reserves;
ng = size(mpc.gen, 1);
                           %% number of on-line gens
%% variable bounds
Rmin = zeros(ng, 1);
                                 %% bound below by 0
                                 \% bound above by stated max reserve qty ...
Rmax = r.qty;
k = find(mpc.gen(:, RAMP_10) > 0 & mpc.gen(:, RAMP_10) < Rmax);</pre>
Rmax = Rmax / mpc.baseMVA;
%% constraints
I = speye(ng);
                                 %% identity matrix
Ar = [I I];
Pmax = mpc.gen(:, PMAX) / mpc.baseMVA;
lreq = r.req / mpc.baseMVA;
%% cost
Cw = r.cost * mpc.baseMVA;
                                %% per unit cost coefficients
%% add them to the model
om.add_var('R', ng, [], Rmin, Rmax);
om.add_lin_constraint('Pg_plus_R', Ar, [], Pmax, {'Pg', 'R'});
om.add_lin_constraint('Rreq', r.zones, lreq, [], {'R'});
om.add_quad_cost('Rcost', [], Cw, 0, {'R'});
```

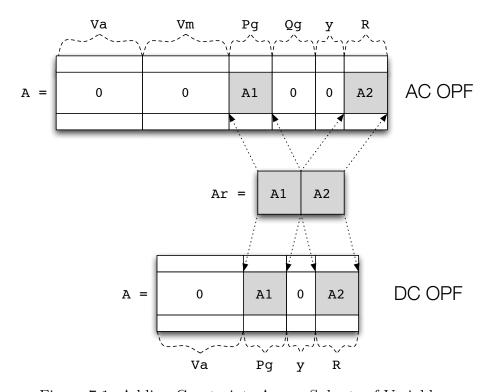


Figure 7-1: Adding Constraints Across Subsets of Variables

7.3.3 int2ext Callback

After the simulation is complete and before the results are printed or saved, MAT-POWER converts the case data in the results struct back to external indexing by calling the following.

```
results = int2ext(results, mpopt);
```

This conversion essentially undoes everything that was done by ext2int. Buses are restored to their original numbering and all out-of-service or isolated generators, branches and buses are restored.

This callback is invoked from int2ext immediately before the resulting case is converted from internal back to external indexing. At this point, the simulation has been completed and the results struct, a superset of the original MATPOWER case struct passed to the OPF, contains all of the results. This results struct is passed to the callback, along with the MATPOWER options struct mpopt,⁵¹ and any (optional) args supplied when the callback was registered via add_userfcn. The output of the callback is the updated results struct. This is typically used to convert any results to external indexing and populate any corresponding fields in the results struct.

The results struct contains, in addition to the standard OPF results, solution information related to all of the user-defined variables, constraints and costs. Table 7-3 summarizes where the various data is found. Each of the fields listed in the table is actually a struct whose fields correspond to the named sets created by add_var, add_lin_constraint, add_nln_constraint, add_quad_cost, add_nln_cost and add_legacy_cost.

In the example code below, the callback function begins by converting the reserves input data in the resulting case (qty, cost and zones fields of results.reserves) back to external indexing via calls to i2e_field. See the help for i2e_field and i2e_data for more details on how they can be used.

⁵¹The mpopt may be empty if int2ext() is called manually without providing it.

Table 7-3: Results for User-Defined Variables, Constraints and Costs

name	description
results.var.val	final value of user-defined variables
results.var.mu.l	shadow price on lower limit of user-defined variables
results.var.mu.u	shadow price on upper limit of user-defined variables
results.lin.mu.l	shadow price on lower (left-hand) limit of linear constraints
results.lin.mu.u	shadow price on upper (right-hand) limit of linear constraints
results.nle.lambda	shadow price on nonlinear equality constraints
results.nli.mu	shadow price on nonlinear inequality constraints
results.cost	final value of legacy user costs
results.nlc	final value of general nonlinear costs
results.qdc	final value of quadratic costs

Then the reserves results of interest are extracted from the appropriate sub-fields of results.var, results.lin and results.cost, converted from per unit to per MW where necessary, and stored with external indexing for the end user in the chosen fields of the results struct.

```
function results = userfcn_reserves_int2ext(results, mpopt, args)
%%---- convert stuff back to external indexing ----
%% convert all reserve parameters (zones, costs, qty, rgens)
results = i2e_field(results, {'reserves', 'qty'}, 'gen');
results = i2e_field(results, {'reserves', 'cost'}, 'gen');
results = i2e_field(results, {'reserves', 'zones'}, 'gen', 2);
r = results.reserves;
ng0 = size(results.order.ext.gen, 1); %% number of gens (external)
%%---- results post-processing -----
%% get the results (per gen reserves, multipliers) with internal gen indexing
%% and convert from p.u. to per MW units
[RO, Rl, Ru] = results.om.params_var('R');
       = results.var.val.R * results.baseMVA;
       = Rl * results.baseMVA;
Rmax = Ru * results.baseMVA;
mu_l = results.var.mu.l.R / results.baseMVA;
     = results.var.mu.u.R / results.baseMVA;
mu_u
mu_Pmax = results.lin.mu.u.Pg_plus_R / results.baseMVA;
%% store in results in results struct
z = zeros(ng0, 1);
results.reserves.R
                     = i2e_data(results, R, z, 'gen');
results.reserves.Rmin = i2e_data(results, Rmin, z, 'gen');
results.reserves.Rmax = i2e_data(results, Rmax, z, 'gen');
results.reserves.mu.l = i2e_data(results, mu_l, z, 'gen');
results.reserves.mu.u = i2e_data(results, mu_u, z, 'gen');
results.reserves.mu.Pmax = i2e_data(results, mu_Pmax, z, 'gen');
results.reserves.prc
                       = z;
for k = 1:ng0
   iz = find(r.zones(:, k));
   results.reserves.prc(k) = sum(results.lin.mu.l.Rreq(iz)) / results.baseMVA;
end
results.reserves.totalcost = results.cost.Rcost;
```

7.3.4 printpf Callback

The pretty-printing of the standard OPF output is done via a call to printpf after the case has been converted back to external indexing. This callback is invoked from within printpf after the pretty-printing of the standard OPF output. Inputs are the results struct, the file descriptor to write to, a MATPOWER options struct, and any (optional) args supplied via add_userfcn. Output is the results struct. This is typically used for any additional pretty-printing of results.

In this example, the out.all flag in the options struct is checked before printing anything. If it is non-zero, the reserve quantities and prices for each unit are printed first, followed by the per-zone summaries. An additional table with reserve limit shadow prices might also be included.

```
function results = userfcn_reserves_printpf(results, fd, mpopt, args)
%% define named indices into data matrices
[GEN_BUS, PG, QG, QMAX, QMIN, VG, MBASE, GEN_STATUS, PMAX, PMIN, ...
   MU_PMAX, MU_PMIN, MU_QMAX, MU_QMIN, PC1, PC2, QC1MIN, QC1MAX, ...
   QC2MIN, QC2MAX, RAMP_AGC, RAMP_10, RAMP_30, RAMP_Q, APF] = idx_gen;
%%---- print results ----
r = results.reserves;
ng = length(r.R);
nrz = size(r.req, 1);
if mpopt.out.all ~= 0
   fprintf(fd, '\n========);
   fprintf(fd, '\n| Reserves
   fprintf(fd, '\n========);
   fprintf(fd, '\n Gen Bus Status Reserves Price');
   fprintf(fd, '\n #
                          (WW)
                    #
                                        ($/MW)');
   fprintf(fd, '\n--- ----');
   for k = 1:ng
      results.gen(k, GEN_STATUS));
      if results.gen(k, GEN_STATUS) > 0 && abs(results.reserves.R(k)) > 1e-6
         fprintf(fd, '%10.2f', results.reserves.R(k));
      else
         fprintf(fd, ' - ');
      fprintf(fd, '%10.2f ', results.reserves.prc(k));
   fprintf(fd, '\n
                                ----');
                                              Total Cost: $%.2f', ...
   fprintf(fd, '\n
                        Total:%10.2f
      sum(results.reserves.R(r.igr)), results.reserves.totalcost);
   fprintf(fd, '\n');
   fprintf(fd, '\nZone Reserves Price ');
   fprintf(fd, '\n #
                            ($/MW) ');
                   (MW)
   fprintf(fd, '\n----');
   for k = 1:nrz
      fprintf(fd, '\n%3d%10.2f%10.2f', k, sum(results.reserves.R(iz)), ...
                results.lin.mu.l.Rreq(k) / results.baseMVA);
   end
   fprintf(fd, '\n');
   %% print binding reserve limit multipliers ...
end
```

7.3.5 savecase Callback

The savecase is used to save a MATPOWER case struct to an M-file, for example, to save the results of an OPF run. The savecase callback is invoked from savecase after printing all of the other data to the file. Inputs are the case struct, the file descriptor to write to, the variable prefix (typically 'mpc.') and any (optional) args supplied via add_userfcn. Output is the case struct. The purpose of this callback is to write any non-standard case struct fields to the case file.

In this example, the zones, req, cost and qty fields of mpc.reserves are written to the M-file. This ensures that a case with reserve data, if it is loaded via loadcase, possibly run, then saved via savecase, will not lose the data in the reserves field. This callback could also include the saving of the output fields if present. The contributed serialize function⁵² can be very useful for this purpose.

⁵²https://www.mathworks.com/matlabcentral/fileexchange/12063-serialize

```
function mpc = userfcn_reserves_savecase(mpc, fd, prefix, args)
%
   mpc = userfcn_reserves_savecase(mpc, fd, mpopt, args)
%
   This is the 'savecase' stage userfcn callback that prints the M-file
%
   code to save the 'reserves' field in the case file. It expects a
   MATPOWER case struct (mpc), a file descriptor and variable prefix
   (usually 'mpc.'). The optional args are not currently used.
r = mpc.reserves;
fprintf(fd, '\n\%\%\---- Reserve Data ----\%\%\\n');
fprintf(fd, '%%%% reserve zones, element i, j is 1 iff gen j is in zone i\n');
fprintf(fd, '%sreserves.zones = [\n', prefix);
template = '';
for i = 1:size(r.zones, 2)
    template = [template, '\t%d'];
end
template = [template, ';\n'];
fprintf(fd, template, r.zones.');
fprintf(fd, '];\n');
fprintf(fd, '\n%%%% reserve requirements for each zone in MW\n');
fprintf(fd, '%sreserves.req = [\t%g', prefix, r.req(1));
if length(r.req) > 1
    fprintf(fd, ';\t%g', r.req(2:end));
end
fprintf(fd, '\t];\n');
fprintf(fd, '\n\%%% reserve costs in $/MW for each gen\n');
fprintf(fd, '%sreserves.cost = [\t%g', prefix, r.cost(1));
if length(r.cost) > 1
    fprintf(fd, ';\t%g', r.cost(2:end));
end
fprintf(fd, '\t];\n');
if isfield(r, 'qty')
    fprintf(fd, '\n\%\% max reserve quantities for each gen\n');
    fprintf(fd, '%sreserves.qty = [\t%g', prefix, r.qty(1));
    if length(r.qty) > 1
        fprintf(fd, ';\t%g', r.qty(2:end));
    fprintf(fd, '\t];\n');
end
%% save output fields for solved case ...
```

7.4 Registering the Callbacks

As seen in the fixed zonal reserve example, adding a single extension to the standard OPF formulation is often best accomplished by a set of callback functions. A typical use case might be to run a given case with and without the reserve requirements active, so a simple method for enabling and disabling the whole set of callbacks as a single unit is needed.

The recommended method is to define all of the callbacks in a single file containing a "toggle" function that registers or removes all of the callbacks depending on whether the value of the second argument is 'on' or 'off'. The state of the registration of any callbacks is stored directly in the mpc struct. In our example, the toggle_reserves.m file contains the toggle_reserves function as well as the five callback functions.

```
function mpc = toggle_reserves(mpc, on_off)
%TOGGLE_RESERVES Enable, disable or check status of fixed reserve requirements.
   MPC = TOGGLE_RESERVES(MPC, 'on')
   MPC = TOGGLE_RESERVES(MPC, 'off')
   T_F = TOGGLE_RESERVES(MPC, 'status')
if strcmp(upper(on_off), 'ON')
    % <code to check for required 'reserves' fields in mpc>
    %% add callback functions
   mpc = add_userfcn(mpc, 'ext2int', @userfcn_reserves_ext2int);
   mpc = add_userfcn(mpc, 'formulation', @userfcn_reserves_formulation);
   mpc = add_userfcn(mpc, 'int2ext', @userfcn_reserves_int2ext);
   mpc = add_userfcn(mpc, 'printpf', @userfcn_reserves_printpf);
   mpc = add_userfcn(mpc, 'savecase', @userfcn_reserves_savecase);
   mpc.userfcn.status.dcline = 1;
elseif strcmp(upper(on_off), 'OFF')
   mpc = remove_userfcn(mpc, 'savecase', @userfcn_reserves_savecase);
   mpc = remove_userfcn(mpc, 'printpf', @userfcn_reserves_printpf);
   mpc = remove_userfcn(mpc, 'int2ext', @userfcn_reserves_int2ext);
   mpc = remove_userfcn(mpc, 'formulation', @userfcn_reserves_formulation);
   mpc = remove_userfcn(mpc, 'ext2int', @userfcn_reserves_ext2int);
   mpc.userfcn.status.dcline = 0;
elseif strcmp(upper(on_off), 'STATUS')
    if isfield(mpc, 'userfcn') && isfield(mpc.userfcn, 'status') && ...
            isfield(mpc.userfcn.status, 'dcline')
       mpc = mpc.userfcn.status.dcline;
       mpc = 0;
    end
else
    error('toggle_dcline: 2nd argument must be ''on'', ''off'' or ''status''');
end
```

Running a case that includes the fixed reserves requirements is as simple as loading the case, turning on reserves and running it.

```
mpc = loadcase('t_case30_userfcns');
mpc = toggle_reserves(mpc, 'on');
results = runopf(mpc);
```

7.5 Summary

The five callback stages currently defined by MATPOWER are summarized in Table 7-4.

Table 7-4: Callback Functions

name	invoked	typical use
ext2int	from ext2int immediately after case data is converted from external to internal indexing.	Check consistency of input data, convert to internal indexing.
formulation	from opf after OPF Model (om) object is initialized with standard OPF formulation.	Modify OPF formulation, by adding user-defined variables, constraints, costs.
int2ext	from int2ext immediately before case data is converted from internal back to external indexing.	Convert data back to external indexing, populate any additional fields in the results struct.
printpf	from printpf after pretty-printing the standard OPF output.	Pretty-print any results not included in standard OPF.
savecase	from savecase after printing all of the other case data to the file.	Write non-standard case struct fields to the case file.

7.6 Example Extensions

MATPOWER includes three OPF extensions implementing via callbacks, respectively, the co-optimization of energy and reserves, interface flow limits and dispatchable DC transmission lines.

7.6.1 Fixed Zonal Reserves

This extension is a more complete version of the example of fixed zonal reserve requirements used for illustration above in Sections 7.3 and 7.4. The details of the extensions to the standard OPF problem are given in equations (7.2)–(7.5) and a description of the relevant input and output data structures is summarized in Tables 7-5 and 7-6, respectively.

The code for implementing the callbacks can be found in toggle_reserves. A wrapper around runopf that turns on this extension before running the OPF is

Table 7-5: Input Data Structures for Fixed Zonal Reserves

name	description	
mpc reserves	MATPOWER case struct additional field in mpc containing input parameters for zonal reserves in the	
	following sub-fields:	
cost	$(n_q \text{ or } n_{qr}^{\dagger}) \times 1 \text{ vector of reserve costs in $/MW, } c \text{ from (7.3)}$	
qty	$(n_g \text{ or } n_{gr}^{\dagger}) \times 1$ vector of reserve quantity upper bounds in MW, i^{th} element is r_i^{max}	
zones	$n_{rz} \times n_g$ matrix of reserve zone definitions	
	$\mathtt{zones(k,j)} = \left\{ egin{array}{ll} 1 & ext{if gen } j ext{ belongs to reserve zone } k & (j \in Z_k) \\ 0 & ext{otherwise} & (j \notin Z_k) \end{array} ight.$	
req	$n_{rz} \times 1$ vector of zonal reserve requirements in MW, k^{th} element is R_k from (7.5)	

[†] Here n_{qr} is the number of generators belonging to at least one reserve zone.

Table 7-6: Output Data Structures for Fixed Zonal Reserves

name	description
results	OPF results struct, superset of mpc with additional fields for output data
reserves	zonal reserve data, with results in the following sub-fields: [‡]
R	reserve allocation r_i , in MW
Rmin	lower bound of r_i in MW from (7.2), i.e. all zeros
Rmax	upper bound of r_i in MW from (7.2), i.e. $\min(r_i^{\max}, \Delta_i)$
mu	shadow prices on constraints in \$/MW
1	shadow price on lower bound of r_i in (7.2)
u	shadow price on upper bound of r_i in (7.2)
Pmax	shadow price on capacity constraint (7.4)
prc	reserve price in \$/MW, computed for unit i as the sum of the shadow prices of constraints k in (7.5) in which unit i participates $(k \mid i \in Z_k)$

[‡] All zonal reserve result sub-fields are $n_g \times 1$ vectors, where the values are set to zero for units not participating in the provision of reserves.

provided in runopf_w_res, allowing you to run a case with an appropriate reserves field, such as t_case30_userfcns, as follows.

```
results = runopf_w_res('t_case30_userfcns');
```

See help runopf_w_res and help toggle_reserves for more information. Examples of using this extension and a case file defining the necessary input data can be found in t_opf_userfcns and t_case30_userfcns, respectively. Additional tests for runopf_w_res are included in t_runopf_w_res.

7.6.2 Interface Flow Limits

This extension adds interface flow limits based on flows computed from a DC network model. It is implemented in toggle_iflims. A flow interface k is defined as a set \mathcal{B}_k of branch indices i and a direction for each branch. If p_i represents the real power flow ("from" bus \rightarrow "to" bus) in branch i and d_i is equal to 1 or -1 to indicate the direction, ⁵³ then the interface flow f_k for interface k is defined as

$$f_k(\Theta) = \sum_{i \in \mathcal{B}_k} d_i p_i(\Theta), \tag{7.6}$$

where each branch flow p_i is an approximation calculated as a linear function of the bus voltage angles based on the DC power flow model from equation (3.29).

This extension adds to the OPF problem a set of $n_{\rm if}$ doubly-bounded constraints on these flows.

$$F_k^{\min} \le f_k(\Theta) \le F_k^{\max} \quad \forall k \in \mathcal{I}_f$$
 (7.7)

where F_k^{\min} and F_k^{\max} are the specified lower and upper bounds on the interface flow, and \mathcal{I}_f is a the set indices of interfaces whose flow limits are to be enforced.

The data for the problem is specified in an additional if field in the MATPOWER case struct mpc. This field is itself a struct with two sub-fields, map and lims, used for input data, and two others, P and mu, used for output data. The format of this data is described in detail in Tables 7-7 and 7-8.

Table 7-7: Input Data Structures for Interface Flow Limits

name	description
mpc	Matpower case struct
if	additional field in mpc containing input parameters for interface flow limits in the following sub-fields:
map	$(\sum_k n_k) \times 2$ matrix defining the interfaces, where n_k is the number branches that belong to interface k . The n_k branches of interface k are defined by
	n_k rows in the matrix, where the first column in each is equal to k and the second is equal to the corresponding branch index i multiplied by d_i to indicate the direction.
lims	$n_{\rm if} \times 3$ matrix of interface limits, where $n_{\rm if}$ is the number of interface limits to be enforced. The first column is the index k of the interface, and the second and third columns are $F_k^{\rm min}$ and $F_k^{\rm max}$, the lower and upper limits respectively, on the DC model flow limits (in MW) for the interface.

⁵³If $d_i = 1$, the definitions of the positive flow direction for the branch and the interface are the same. If $d_i = -1$, they are opposite.

Table 7-8: Output Data Structures for Interface Flow Limits

name	description
results if	OPF results struct, superset of mpc with additional fields for output data additional field in results containing output parameters for interface flow limits in the following sub-fields:
P	$n_{\rm if} \times 1$ vector of actual flow in MW across the corresponding interface (as measured at the "from" end of associated branches)
mu.l mu.u	$n_{\rm if} \times 1$ vector of shadow prices on lower flow limits $(u/{\rm MW})^{\dagger}$ $n_{\rm if} \times 1$ vector of shadow prices on upper flow limits $(u/{\rm MW})^{\dagger}$

^{\dagger} Here we assume the objective function has units u.

See help toggle_iflims for more information. Examples of using this extension and a case file defining the necessary input data for it can be found in t_opf_userfcns and t_case30_userfcns, respectively. Note that, while this extension can be used for AC OPF problems, the actual AC interface flows will not necessarily be limited to the specified values, since it is a DC flow approximation that is used for the constraint.

Running a case that includes the interface flow limits is as simple as loading the case, turning on the extension and running it. Unlike with the reserves extension, MATPOWER does not currently have a wrapper function to automate this.

```
mpc = loadcase('t_case30_userfcns');
mpc = toggle_iflims(mpc, 'on');
results = runopf(mpc);
```

7.6.3 DC Transmission Lines

Beginning with version 4.1, MATPOWER also includes a simple model for dispatchable DC transmission lines. While the implementation is based on the extensible OPF architecture described above, it can be used for simple power flow problems as well, in which the case the (OPF only) formulation callback is skipped.

A DC line in MATPOWER is modeled as two linked "dummy" generators, as shown in Figures 7-2 and 7-3, one with negative capacity extracting real power from the network at the "from" end of the line and another with positive capacity injecting power into the network at the "to" end. These dummy generators are added by the ext2int callback and removed by the int2ext callback. The real power flow p_f on the DC line at the "from" end is defined to be equal to the negative of the injection of corresponding dummy generator. The flow at the "to" end p_t is defined to be equal to the injection of the corresponding generator.

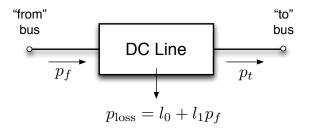


Figure 7-2: DC Line Model

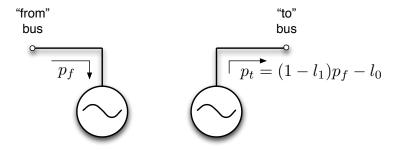


Figure 7-3: Equivalent "Dummy" Generators

MATPOWER links the values of p_f and p_t using the following relationship, which includes a linear approximation of the real power loss in the line.

$$p_{t} = p_{f} - p_{loss}$$

$$= p_{f} - (l_{0} + l_{1}p_{f})$$

$$= (1 - l_{1})p_{f} - l_{0}$$
(7.8)

Here the linear coefficient l_1 is assumed to be a small ($\ll 1$) positive number. Obviously, this is not applicable for bi-directional lines, where the flow could go either direction, resulting in decreasing losses for increasing flow in the "to" \rightarrow "from" direction. There are currently two options for handling bi-directional lines. The first is to use a constant loss model by setting $l_1 = 0$. The second option is to create two separate but identical lines oriented in opposite directions. In this case, it is important that the lower limit on the flow and the constant term of the loss model l_0 be set to zero to ensure that only one of the two lines has non-zero flow at a time.⁵⁴

Upper and lower bounds on the value of the flow can be specified for each DC line, along with an optional operating cost. It is also assumed that the terminals of the line have a range of reactive power capability that can be used to maintain a voltage setpoint. Just as with a normal generator, the voltage setpoint is only used for simple power flow; the OPF dispatches the voltage anywhere between the lower and upper bounds specified for the bus. Similarly, in a simple power flow the input value for p_f and the corresponding value for p_t , computed from (7.8), are used to specify the flow in the line.

Most of the data for DC lines is stored in a dcline field in the MATPOWER case struct mpc. This field is a matrix similar to the branch matrix, where each row corresponds to a particular DC line. The columns of the matrix are defined in Table B-5 and include connection bus indices, line status, flows, terminal reactive injections, voltage setpoints, limits on power flow and VAr injections, and loss parameters. Also, similar to the branch or gen matrices, some of the columns are used for input values, some for results, and some, such as PF can be either input or output, depending on whether the problem is a simple power flow or an optimal power flow. The idx_dcline function defines a set of constants for use as named column indices for the dcline matrix.

An optional dclinecost matrix, in the same form as gencost, can be used to specify a cost to be applied to p_f in the OPF. If the dclinecost field is not present, the cost is assumed to be zero.

⁵⁴A future version may make the handling of this second option automatic.

MATPOWER'S DC line handling is implemented in toggle_dcline and examples of using it can be found in t_dcline. The case file t_case9_dcline includes some example DC line data. See help toggle_dcline for more information.

Running a case that includes DC lines is as simple as loading the case, turning on the extension and running it. Unlike with the reserves extension, MATPOWER does not currently have a wrapper function to automate this.

```
mpc = loadcase('t_case9_dcline');
mpc = toggle_dcline(mpc, 'on');
results = runopf(mpc);
```

7.6.4 OPF Soft Limits

MATPOWER includes an extension that replaces limits (such as branch flow limits, voltage magnitude bounds, generator limits, etc.) in an optimal power flow with soft limits, that is, limits that can be violated with some linear penalty cost. This can be useful in identifying the cause of infeasibility in some optimal power flow problems. The extension is implemented in toggle_softlims. A limited version for branch flow constraints in a DC model only, was introduced in MATPOWER 5.0, with a generalization to all of the standard OPF limits added in MATPOWER 7.0. The soft AC branch flow limit implemention provides an example of user-defined nonlinear constraints.

In general, replacing a hard constraint of the form

$$h^i(x) \le 0 \tag{7.9}$$

with a soft limit involves introducing a new non-negative variable s_i to represent the violation of the original constraint. The original hard constraint (7.9) is replaced by one in which this violation variable is used to "relax" the constraint by adding it to the right-hand.

$$s^i \ge 0 \tag{7.10}$$

$$h^i(x) \le s^i. \tag{7.11}$$

Finally, a penalty cost on violations is implemented via a linear cost coefficient, c_{v}^{i} , applied to each violation variable s^{i} , so that the additional user defined cost term from (6.47) for the set of soft limits looks like

$$f_u(\hat{x}) = c_v^{\mathsf{T}} s. \tag{7.12}$$

Take, as an example, branch flow limits. The flow constraints in (6.9) and (6.10)are rewritten by replacing the right hand side with a flow violation variable $s_{\mathtt{rate-a}}^i$

$$s_{\text{rate a}}^i \ge 0 \qquad \forall i \qquad (7.13)$$

$$s_{\mathtt{rate_a}}^{i} \ge 0 \qquad \forall i \qquad (7.13)$$

$$h_{f}^{i}(\Theta, V_{m}) = \left| f_{f}^{i}(\Theta, V_{m}) \right| - f^{i, \max} \le s_{\mathtt{rate_a}}^{i} \qquad \forall i \qquad (7.14)$$

$$h_t^i(\Theta, V_m) = \left| f_t^i(\Theta, V_m) \right| - f^{i, \max} \le s_{\mathtt{rate}, \mathtt{a}}^i \quad \forall i. \tag{7.15}$$

In the case of AC flow constraints, the flow functions f_f^i and f_t^i are nonlinear and can be either apparent power, active power, or current as shown in (6.11). In the case of DC flow constraints, where there are no losses, these flows can be implemented in terms of the flow at the from end of the line and its negative, both linear functions of the bus voltage angles, based on (6.30) and (6.31).

$$s_{\text{rate a}}^i \ge 0 \qquad \forall i$$
 (7.16)

$$B_f^i \Theta + p_{f,\text{shift}}^i - p_f^{i,\text{max}} \le s_{\text{rate.a}}^i \quad \forall i$$
 (7.17)

$$s_{\mathtt{rate_a}}^{i} \geq 0 \qquad \forall i \qquad (7.16)$$

$$B_{f}^{i}\Theta + p_{f, \mathrm{shift}}^{i} - p_{f}^{i, \mathrm{max}} \leq s_{\mathtt{rate_a}}^{i} \qquad \forall i \qquad (7.17)$$

$$-B_{f}^{i}\Theta - p_{f, \mathrm{shift}}^{i} - p_{f}^{i, \mathrm{max}} \leq s_{\mathtt{rate_a}}^{i} \qquad \forall i. \qquad (7.18)$$

The feasible area for the DC case is illustrated in Figure 7-4.

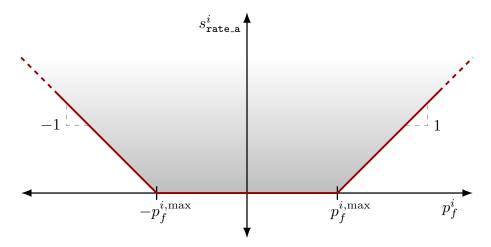


Figure 7-4: Feasible Region for Branch Flow Violation Constraints

MATPOWER implements soft limits for each of the types of constraints listed in Table 7-9, where the formulation for each constraint type is shown.

The parameters for the problem are specified in an additional softlims field in the MATPOWER case struct mpc. This field is itself a struct where each sub-field

Table 7-9: Soft Limit Formulation

name	hard-limit formulation	soft-limit formulation
VMIN, VMAX†	$v_m^i \ge v_m^{i,\min}, v_m^i \le v_m^{i,\max}$	$v_m^i + s_{\mathtt{vmin}}^i \ge v_m^{i,\min}, v_m^i - s_{\mathtt{vmax}}^i \le v_m^{i,\max}$
$\mathtt{RATE}_\mathtt{A}^\ddagger$	$\left f_f^i(\Theta, V_m) \right \le f_{\max}^i$	$\left f_f^i(\Theta,V_m) ight -s_{ exttt{rate_a}} \leq f_{ exttt{max}}^i$
	$\left f_t^i(\Theta, V_m) \right \le f_{\max}^i$	$\left f_t^i(\Theta, V_m)\right - s_{\mathtt{rate_a}} \leq f_{\max}^i$
PMIN, PMAX	$p_g^i \ge p_g^{i,\min}, p_g^i \le p_g^{i,\max}$	$p_g^i + s_{\texttt{pmin}} \ge p_g^{i, \min}, p_g^i - s_{\texttt{pmax}} \le p_g^{i, \max}$
QMIN, QMAX [†]	$q_q^i \ge q_q^{i,\min}, q_q^i \le q_q^{i,\max}$	$q_q^i + s_{ extsf{qmin}} \geq q_q^{i, ext{min}}, q_q^i - s_{ extsf{qmax}} \leq q_q^{i, ext{max}}$
ANGMIN	$ heta_f^i - heta_t^i \geq \Delta heta^{i, \min}$	$ heta_f^i - heta_t^i + s_{\mathtt{angmin}} \geq \Delta heta^{i, \min}$
ANGMAX	$\begin{array}{l} q_g^i \geq q_g^{i,\min}, q_g^i \leq q_g^{i,\max} \\ \theta_f^i - \theta_t^i \geq \Delta \theta^{i,\min} \\ \theta_f^i - \theta_t^i \leq \Delta \theta^{i,\max} \end{array}$	$\begin{aligned} p_g^i + s_{\text{pmin}} &\geq p_g^{i,\text{min}}, p_g^i - s_{\text{pmax}} \leq p_g^{i,\text{max}} \\ q_g^i + s_{\text{qmin}} &\geq q_g^{i,\text{min}}, q_g^i - s_{\text{qmax}} \leq q_g^{i,\text{max}} \\ \theta_f^i - \theta_t^i + s_{\text{angmin}} &\geq \Delta \theta^{i,\text{min}} \\ \theta_f^i - \theta_t^i - s_{\text{angmax}} &\leq \Delta \theta^{i,\text{max}} \end{aligned}$

[†] Only applicable for the AC OPF.

corresponds to the name of one of the limits listed in Table 7-9. For each of the limits, the input parameters are specified in the fields detailed in Table 7-10, with defaults summarized in Table 7-11.

Depending on whether the original constraint in (7.10) is an upper bound or a lower bound, it can be written as one of the following

$$h^{i}(x) = \begin{cases} \bar{h}^{i}(x) - h^{i, \text{lim}} & \text{(upper bound)} \\ h^{i, \text{lim}} - \bar{h}^{i}(x) & \text{(lower bound)} \end{cases}$$
 (7.19)

where $h^{i,\lim}$ is the actual value of the hard limit. When introducing a soft limit, this original hard limit can be removed completely or modified to have a new value. MATPOWER provides a several ways of modifying the original hard limit based on fields hl_mod ("hard limit modification") and hl_val ("hard limit value") in the respective limit structures. This new hard limit is implemented via an upper bound on the corresponding violation variable s^i . Table 7-12 summarizes the new hard limits and the corresponding bounds on the violation variable as functions of the values of hl_mod and hl_val .

Unspecified limits in the mpc.softlims input parameters are handled differently, depending on the value of the MATPOWER option 'opf.softlims.default' (see Table C-5).

- If the opf.softlims.default option is 1 (default), each limit not explicitly specified, whether missing in mpc.softlims or present but empty, will be initialized as in Table 7-11.
- If the opf.softlims.default option is 0, limits that are missing in mpc.softlims are treated as though specified with hl mod = 'none', that is, they are not

[‡] Example of a user-defined nonlinear constraint in the AC version.

Table 7-10: Input Data Structures for OPF Soft Limits

name	description		
mpc	Matpower case struct		
softlims	additional field in mpc containing OPF soft limit input parameters for the possible limits, each of which is optional.*		
<lim></lim>	<lim> refers to the name of the limit, i.e. one of the following: ANGMIN, ANGMAX, RATE_A, PMIN, PMAX, QMIN, QMAX, VMIN, VMAX. Each of these is a struct with input parameters defining the soft limits for this type of constraint in the following optional fields (see Table 7-11 for defaults):</lim>		
idx	$n_{\rm sl} \times 1$ vector of row indices for bus, branch or gen matrix, depending on <lim>, specifying the elements for which corresponding soft limits are to be applied[†]</lim>		
busnum [‡]	$n_{\rm sl} \times 1$ vector of external bus numbers specifying the buses to which soft limits are to be applied [†]		
cost	scalar or $n_{\rm sl} \times 1$ vector $c_{\rm v}$ of linear cost coefficients for limit violation costs		
$\mathtt{hl_mod}^\S$	string indicating type of modification to original hard limit:		
	'none' - do <i>not</i> add soft limit, no change to original hard limit		
	'remove' - add soft limit, relax hard limit by removing it completely		
	'replace' - add soft limit, relax hard limit by replacing original with value specified in hl_val		
	'scale' - add soft limit, relax hard limit by scaling original by value specified in hl_val		
	'shift' - add soft limit, relax hard limit by shifting original by value specified in hl_val		
$\mathtt{hl_val}^{\S}$	scalar or $n_{\rm sl} \times 1$ vector value used to modify hard limit according to hl_mod.		
	Ignored for 'none' and 'remove', required for 'replace', and optional, with		
	the following defaults, for 'scale' and 'shift':		
	'scale' -2 for positive upper limits or negative lower limits, 0.5 otherwise 'shift' -0.25 for VMAX and VMIN, 10 otherwise		
	SHIII 0.20 for VIAN and VIIIN, 10 Other Wise		

^{*} For fields not present, the value of the 'opf.softlims.defaults' option determines whether the corresponding limit is unchanged (= 0) or converted (= 1, default) to a soft limit using default values from Table 7-11.

converted to soft limits at all. On the other hand, the default soft limits of Table 7-11 are still applied to any limit that is present in mpc.softlims but empty. For example, mpc.softlims.RATE_A = struct() would result in the default soft branch limits being applied.

Soft limit outputs, summarized in Table 7-13, include the amount and cost of any overloads, that is, violations of the original hard limits. These can be found as additional fields, overload and ov_cost, under each limit in results.softlims. The Kuhn-Tucker multipliers on the soft limit constraints are also included in the

 $^{^\}dagger$ For bus constraints, idx overrides $\tt busnum$ if both are provided.

[‡] Only applicable for bus constraints, i.e. <LIM> is VMIN or VMAX.

[§] Any new, relaxed hard limits are implemented via bounds on the violation variables. See Table 7-12 for details.

Table 7-11: Default Soft Limit Values

name	description
VMIN	voltage magnitude lower bound
idx	all buses
cost	100,000/p.u. [†]
hl_mod	'replace'
$\mathtt{hl_val}$	0
VMAX	voltage magnitude upper bound
idx	all buses
cost	$100,000/\text{p.u.}^{\dagger}$
hl_mod	'remove'
$RATE_A$	branch flow limit
idx	all on-line branches with bounded flow limits
cost	$1000/MVA$, $1000/MW$, or $1000/(kA \cdot V_{basekV})^{\dagger}$ where the units depend on 'opf.flow_lim'
hl_mod	'remove'
PMIN	generator active power lower bound
idx	all on-line generators, excluding dispatchable loads
cost	$1000/\mathrm{MW}^\dagger$
hl_mod	'replace'
$\mathtt{hl_val}$	0 if $P_q^{i,\min} \geq 0, -\infty$ otherwise
PMAX, QMIN, QMAX	generator active power upper bound, reactive power lower/upper bounds
idx	all on-line generators, excluding dispatchable loads
cost	$1000/MW \text{ or } 1000/MVAr^{\dagger}$
hl_mod	'remove'
ANGMIN, ANGMAX	branch angle difference lower/upper bounds
idx	all on-line branches with angle difference limits, $\Delta\theta$, where $ \Delta\theta < 360^{\circ}$
cost	$1000/\deg^{\dagger}$
hl_mod	'remove'

[†] Unless the maximum marginal cost at P^{\max} across all online generators, which we call c_g^{\max} , exceeds \$1000/MWh, in which case the numerical value of the default cost is c_g^{\max} instead of 1000 (or $100 \cdot c_g^{\max}$ instead of 100,000, for VMIN or VMAX).

usual columns of the corresponding matrix in the results, e.g. bus(:, MU_VMAX), branch(:, MU_SF), etc. When there are no violations, these shadow prices are the same as if hard limits were imposed. When there is a violation, they are equal to the cost $c_{\rm v}$ associated with the violation variable, except in the case where a modified hard limit is also binding, in which case it also reflects the shadow price on this modified hard constraint.

Table 7-12: Possible Hard-Limit Modifications

h_mod	new hard limit $h_{\text{new}}^{i,\text{lim}}$	upper bound on violation variable s^i
'remove'	$\pm\infty^{\dagger}$	∞
'replace'	$\mathtt{hl}_\mathtt{val}$	$egin{aligned} \left ext{hl_val} - ext{h}_{ ext{orig}}^{ ext{i,lim}} ight \ \left ext{(hl_val} - ext{1}) \cdot ext{h}_{ ext{orig}}^{ ext{i,lim}} ight \end{aligned}$
'scale'	$\mathtt{hl_val} \cdot \mathtt{h}^{\mathtt{i}, \mathrm{lim}}_{\mathrm{orig}}$	$\left (\mathtt{hl_val} - \mathtt{1}) \cdot \mathtt{h}^{\mathtt{i}, \mathrm{lim}}_{\mathrm{orig}} \right $
'shift'	$h_{ ext{orig}}^{i, ext{lim}} \pm \mathtt{hl_val}^\dagger$	hl_val

[†] Plus sign for upper bounds and minus sign for lower bounds.

Table 7-13: Output Data Structures for OPF Soft Limits

name	description
results	Matpower case struct
softlims	additional field in results containing OPF soft limit outputs for each limit type
<lim></lim>	<pre><lim> refers to the name of the limit, i.e. one of the following: ANGMIN, ANGMAX, RATE_A, PMIN, PMAX, QMIN, QMAX, VMIN, VMAX. Each</lim></pre>
	of these is a struct with the following output fields:
$ ext{overload}^\dagger$	$n_{b/l/g} \times 1$ vector of overload quantities for this limit (amount of
	original hard limit violation)
${\sf ov_cost}^\dagger$	$n_{b/l/g} \times 1$ vector of overload penalty costs for this limit
<pre>branch(:, MU_ANGMIN)</pre>	Kuhn-Tucker multipliers on angle difference lower bounds [‡]
<pre>branch(:, MU_ANGMAX)</pre>	Kuhn-Tucker multipliers on angle difference upper bounds [‡]
<pre>branch(:, MU_SF)</pre>	Kuhn-Tucker multipliers on branch flow limits (from end) [‡]
<pre>branch(:, MU_ST)</pre>	Kuhn-Tucker multipliers on branch flow limits $(to \text{ end})^{\ddagger}$
gen(:, MU_PMIN)	Kuhn-Tucker multipliers on active gen lower bounds [‡]
gen(:, MU_PMAX)	Kuhn-Tucker multipliers on active gen upper bounds [‡]
gen(:, MU_QMIN)	Kuhn-Tucker multipliers on reactive gen lower bounds [‡]
gen(:, MU_QMAX)	Kuhn-Tucker multipliers on reactive gen upper bounds [‡]
bus(:, MU_VMIN)	Kuhn-Tucker multipliers on voltage magnitude lower bounds [‡]
bus(:, MU_VMAX)	Kuhn-Tucker multipliers on voltage magnitude upper bounds [‡]

[†] The dimensions of overload and ov_cost correspond to all buses, branches or generators (depending on <LIM>), not just those whose limits were converted to soft limits. Entries corresponding to those not included (implicitly or explicitly) in idx are set to zero.

[†] For limits that have been converted to soft limits, these are the shadow prices on the soft limit constraints. When there is no violation of the soft limit, this shadow price is the same as it would be for the original hard limit. When there is a violation, it is equal to the corresponding user-supplied violation cost $c_{\mathbf{v}}^{i}$, unless an updated hard constraint is also binding.

See help toggle_softlims for more information on this extension. Examples of using this extension can be found in t_opf_softlims. Running a case that includes the default soft limits is as simple as loading the case, turning on the extension and running it. Unlike with the reserves extension, MATPOWER does not currently have a wrapper function to automate this.

```
mpc = loadcase('case2383wp');
mpc = toggle_softlims(mpc, 'on');
results = runopf(mpc, mpopt);
```

8 Unit De-commitment Algorithm

The standard OPF formulation described in the previous section has no mechanism for completely shutting down generators which are very expensive to operate. Instead they are simply dispatched at their minimum generation limits. MATPOWER includes the capability to run an optimal power flow combined with a unit de-commitment for a single time period, which allows it to shut down these expensive units and find a least cost commitment and dispatch. To run this for case30, for example, type:

```
>> runuopf('case30')
```

By default, runuopf is based on the AC optimal power flow problem. To run a DC OPF, the model option must be set to 'DC'. For convenience, MATPOWER provides a function runduopf which is simply a wrapper that sets the model option to 'DC' before calling runuopf.

MATPOWER uses an algorithm similar to dynamic programming to handle the de-commitment. It proceeds through a sequence of stages, where stage N has N generators shut down, starting with N=0, as follows:

- **Step 1:** Begin at stage zero (N = 0), assuming all generators are on-line with all limits in place.
- Step 2: If the sum of the minimum generation limits for all on-line generators is less than the total system demand, then go to Step 3. Otherwise, go to the next stage, N = N + 1, shut down the generator whose average per-MW cost of operating at its minimum generation limit is greatest and repeat Step 2.
- **Step 3:** Solve a normal OPF. Save the solution as the current best.
- **Step 4:** Go to the next stage, N = N + 1. Using the best solution from the previous stage as the base case for this stage, form a candidate list of generators with minimum generation limits binding. If there are no candidates, skip to Step 6.
- **Step 5:** For each generator on the candidate list, solve an OPF to find the total system cost with this generator shut down. Replace the current best solution with this one if it has a lower cost. If any of the candidate solutions produced an improvement, return to Step 4.
- **Step 6:** Return the current best solution as the final solution.

It should be noted that the method employed here is simply a heuristic. It does not guarantee that the least cost commitment of generators will be found. It is also rather computationally expensive for larger systems and was implemented as a simple way to allow an OPF-based "smart-market", such as described in Appendix F, the option to reject expensive offers while respecting the minimum generation limits on generators.

9 Miscellaneous Matpower Functions

This section describes a number of additional MATPOWER functions that users may find useful. The descriptions here are simply brief summaries, so please use the MATLAB help function to get the full details on each function.

9.1 Input/Output Functions

9.1.1 loadcase

```
mpc = loadcase(casefile)
```

The loadcase function provides the canonical way of loading a MATPOWER case from a file or struct. It takes as input either a struct or the name of an M-file or MAT-file in the MATLAB path (casefile) and returns a standard MATPOWER case struct (mpc). It can also convert from the older version 1 case file format to the current format. This function allows a case to be loaded, and potentially modified, before calling one of the main simulation functions such as runpf or runopf.

9.1.2 savecase

```
savecase(fname, mpc)
savecase(fname, mpc, version)
savecase(fname, comment, mpc)
savecase(fname, comment, mpc, version)
fname = savecase(fname, ...)
```

The savecase function writes out a MATPOWER case file, given a name for the file to be created or overwritten (fname), and a MATPOWER case struct (mpc). If fname ends with '.mat' it saves the case as a MAT-file, otherwise it saves it as an M-file. Optionally returns the filename, with extension added if necessary. The optional comment argument is either string (single line comment) or a cell array of strings which are inserted as comments in the help section of the file. If the optional version argument is '1' it will modify the data matrices to version 1 format before saving.

9.1.3 cdf2mpc

```
mpc = cdf2mpc(cdf_file_name)
mpc = cdf2mpc(cdf_file_name, verbose)
mpc = cdf2mpc(cdf_file_name, mpc_name)
mpc = cdf2mpc(cdf_file_name, mpc_name, verbose)
[mpc, warnings] = cdf2mpc(cdf_file_name, ...)
```

The cdf2mpc function converts an IEEE Common Data Format (CDF) data file into a MATPOWER case struct. Given an optional file name mpc_name, it can save the converted case to a MATPOWER case file. Warnings generated during the conversion process can be optionally returned in the warnings argument.

Since the IEEE CDF format does not contain all of the data needed to run an optimal power flow, some data, such as voltage limits, generator limits and generator costs are created by cdf2mpc. See help cdf2mpc for details.

9.1.4 psse2mpc

```
mpc = psse2mpc(rawfile_name)
mpc = psse2mpc(rawfile_name, verbose)
mpc = psse2mpc(rawfile_name, verbose, rev)
mpc = psse2mpc(rawfile_name, mpc_name)
mpc = psse2mpc(rawfile_name, mpc_name, verbose)
mpc = psse2mpc(rawfile_name, mpc_name, verbose, rev)
[mpc, warnings] = psse2mpc(rawfile_name, ...)
```

The psse2mpc function converts a PSS/E RAW data file into a MATPOWER case struct. Given an optional file name mpc_name, it can save the converted case to a MATPOWER case file. Warnings generated during the conversion process can be optionally returned in the warnings argument. By default, psse2mpc attempts to determine the revision of the PSS/E RAW file from the contents, but the user can specify an explicit revision number to use in the optional rev argument.

9.1.5 save2psse

```
save2psse(fname, mpc)
fname_out = save2psse(fname, mpc)
```

The save2psse function saves a MATPOWER case struct mpc as a PSS/E RAW file. The fname parameter is a string containing the name of the file to be created or overwritten. If fname does not include a file extension, codeq.raw will be added. Optionally returns the, possibly updated, filename. Currently exports to RAW format Rev 33.

9.2 System Information

9.2.1 case_info

```
case_info(mpc)
case_info(mpc, fd)
[groups, isolated] = case_info(mpc)
```

The case_info function prints out detailed information about a MATPOWER case, including connectivity information, summarizing the generation, load and other data by interconnected island. It can optionally print the output to an open file, whose file identifier (as returned by fopen) is specified in the optional second parameter fd. Optional return arguments include groups and isolated buses, as returned by the find islands function.

9.2.2 compare_case

```
compare_case(mpc1, mpc2)
```

Compares the bus, branch and gen matrices of two MATPOWER cases and prints a summary of the differences. For each column of the matrix it prints the maximum of any non-zero differences.

9.2.3find_islands

```
groups = find_islands(mpc)
[groups, isolated] = find_islands(mpc)
```

The find_islands function returns the islands in a network. The return value groups is a cell array of vectors of the bus indices for each island. The second and optional return value isolated is a vector of indices of isolated buses that have no connecting branches.

9.2.4get_losses

```
loss = get_losses(results)
loss = get_losses(baseMVA, bus, branch)
[loss, chg] = get_losses(results)
[loss, fchg, tchg] = get_losses(results)
[loss, fchg, tchg, dloss_dv] = get_losses(results)
[loss, fchg, tchg, dloss_dv, dchg_dvm] = get_losses(results)
```

The get_losses function computes branch series losses, and optionally reactive injections from line charging, as functions of bus voltages and branch parameters, using the following formulae for a branch, as described in Section 3.2, connecting bus f to bus t:

$$loss_{i} = \frac{\left|\frac{v_{f}}{\tau e^{j\theta_{\text{shift}}}} - v_{t}\right|^{2}}{r_{s} - jx_{s}}$$

$$f_{\text{chg}} = \left|\frac{v_{f}}{\tau e^{j\theta_{\text{shift}}}}\right|^{2} \frac{b_{c}}{2}$$
(9.1)

$$f_{\rm chg} = \left| \frac{v_f}{\tau e^{j\theta_{\rm shift}}} \right|^2 \frac{b_c}{2} \tag{9.2}$$

$$t_{\rm chg} = |v_t|^2 \frac{b_c}{2} \tag{9.3}$$

It can also optionally compute the partial derivatives of the line losses and reactive charging injections with respect to voltage angles and magnitudes.

9.2.5 margcost

```
marginalcost = margcost(gencost, Pg)
```

The margcost function computes the marginal cost for generators given a matrix in gencost format and a column vector or matrix of generation levels. The return value has the same dimensions as Pg. Each row of gencost is used to evaluate the cost at the output levels specified in the corresponding row of Pg. The rows of gencost can specify either polynomial or piecewise linear costs and need not be uniform.

9.2.6 isload

```
TorF = isload(gen)
```

The isload function returns a column vector of 1's and 0's. The 1's correspond to rows of the gen matrix which represent dispatchable loads. The current test is $P_{\min} < 0$ and $P_{\max} = 0$.

9.2.7 loadshed

```
shed = loadshed(gen)
shed = loadshed(gen, ild)
```

The loadshed function returns a column vector of MW curtailments of dispatchable loads, computed as the difference between the PG and PMIN values in the corresponding rows of the gen matrix. The optional ild argument is a column vector of generator indices to the dispatchable loads of interest.

9.2.8 printpf

```
printpf(results, fd, mpopt)
```

The printpf function prints power flow and optimal power flow results, as returned to fd, a file identifier which defaults to STDOUT (the screen). The details of what gets printed are controlled by an optional MATPOWER options struct mpopt.

9.2.9 total_load

```
Pd = total_load(mpc)
Pd = total_load(mpc, load_zone, opt, mpopt)
Pd = total_load(bus)
Pd = total_load(bus, gen, load_zone, opt, mpopt)
[Pd, Qd] = total_load(...)
```

The total_load function returns a vector of total load in each load zone. The opt argument controls whether it includes fixed loads, dispatchable loads or both, and for dispatchable loads, whether to use the nominal or realized load values. The load_zone argument defines the load zones across which loads will be summed. It uses the BUS_AREA column (7) of the bus matrix by default. The string value 'all' can be used to specify a single zone including the entire system. The reactive demands are also optionally available as an output.

9.2.10 totcost

```
totalcost = totcost(gencost, Pg)
```

The totcost function computes the total cost for generators given a matrix in gencost format and a column vector or matrix of generation levels. The return value has the same dimensions as Pg. Each row of gencost is used to evaluate the cost at the output levels specified in the corresponding row of Pg. The rows of gencost can specify either polynomial or piecewise linear costs and need not be uniform.

9.3 Modifying a Case

9.3.1 extract_islands

```
mpc_array = extract_islands(mpc)
mpc_array = extract_islands(mpc, groups)
mpc_k = extract_islands(mpc, k)
mpc_k = extract_islands(mpc, groups, k)
mpc_k = extract_islands(mpc, k, custom)
mpc_k = extract_islands(mpc, groups, k, custom)
```

The extract_islands function extracts individual islands in a network that is not fully connected. The original network is specified as a MATPOWER case struct (mpc) and the result is returned as a cell array of case structs, or as a single case struct. Supplying the optional group avoids the need to traverse the network again, saving

time on large systems. A final optional argument custom is a struct that can be used to indicate custom fields of mpc from which to extract data corresponding to buses generators, branches or DC lines.

9.3.2 load2disp

```
mpc = load2disp(mpc0)
mpc = load2disp(mpc0, fname)
mpc = load2disp(mpc0, fname, idx)
mpc = load2disp(mpc0, fname, idx, voll)
```

The load2disp function takes a MATPOWER case mpc0, converts fixed loads to dispatchable loads, curtailable at a specific price, and returns the resulting case struct mpc. It can optionally save the resulting case to a file (fname), convert loads only at specific buses (idx), and set the value of lost load (vol1) to be used as the curtailment price (default is \$5,000/MWh).

9.3.3 modcost

```
newgencost = modcost(gencost, alpha)
newgencost = modcost(gencost, alpha, modtype)
```

The modcost function can be used to modify generator cost functions by shifting or scaling them, either horizontally or vertically. The alpha argument specifies the numerical value of the modification, and modtype defines the type of modification as a string that takes one of the following values: 'SCALE_F' (default), 'SCALE_X', 'SHIFT_F', or 'SHIFT_X'.

9.3.4 scale_load

```
mpc = scale_load(load, mpc)
mpc = scale_load(load, mpc, load_zone)
mpc = scale_load(load, mpc, load_zone, opt)
bus = scale_load(load, bus)
[bus, gen] = scale_load(load, bus, gen, load_zone, opt)
[bus, gen, gencost] = ...
scale_load(load, bus, gen, load_zone, opt, gencost)
```

The scale_load function is used to scale active (and optionally reactive) loads in each zone by a zone-specific ratio, i.e. R(k) for zone k. The amount of scaling for

each zone, either as a direct scale factor or as a target quantity, is specified in load. The load zones are defined by load_zone, and opt specifies the type of scaling (factor or target quantity) and which loads are affected (active, reactive or both and fixed, dispatchable or both). The costs (gencost) associated with dispatchable loads can also be optionally scaled with the loads.

9.3.5 apply_changes

```
mpc_modified = apply_changes(label, mpc_original, chgtab)
```

The apply_changes function implements a general mechanism to apply a set of changes to a base MATPOWER case. This can be used, for example, to define and apply a set of contingencies. There are three basic types of changes, those that replace old values with new ones, those that scale old values by some factor, and those that add a constant to existing values.

The *change table* matrix, chgtab, specifies modifications to be applied to an existing case. These modifications are grouped into sets, designated *change sets*, that are always applied as a group. A change set consists of one or more changes, each specified in a separate row in the chgtab, where the rows share a common label (integer ID).

For example, n_c change sets can be used to define n_c contingencies via a single chgtab with many rows, but only n_c unique labels. The chgtab also optionally specifies a probability π_k of occurance associated with change set k. Table 9-1 summarizes the meaning of the data in each column of chgtab. All of the names referenced in Tables 9-1 through 9-4 are defined as constants by the idx_ct function. Type help idx_ct at the MATLAB prompt for more details. Use of the named constants when constructing a chgtab matrix is encouraged to improve readability.

The value in the CT_TABLE column of chgtab defines which data table is to be modified and the options are given in Table 9-2. With the exception of load and certain generator cost changes, each individual change record specifies modification(s) to a single column of a particular data matrix, either bus, gen, branch or gencost. Some are changes to that column for an individual row in the matrix (or all rows, if the row index is set to 0), while others are area-wide changes that modify all rows corresponding to the specified area.⁵⁵

Load changes are special and may modify multiple columns of the bus and/or gen tables. They offer a more flexible and convenient means of specifying modifications

 $^{^{55}}$ Areas are defined by the BUS_AREA column of the bus matrix.

Table 9-1: Columns of chgtab

name	column	description
CT_LABEL	1	change set label, unique for each change set (integer)
CT_PROB	2	change set probability (number between 0 and 1) †
CT_TABLE	3	type of table to be modified (see Table 9-2 for possible values)
$\mathtt{CT}_\mathtt{ROW}$	4	row index of data to be modified, 0 means all rows,
		for area-wide changes this is the area index, rather than row index
$\mathtt{CT}_{-}\mathtt{COL}$	5	column index of data to be modified (see Table 9-4 for exceptions)
CTCHGTYPE	6	type of change, e.g. replace, scale or add (see Table 9-3 for details)
$\mathtt{CT_NEWVAL}$	7	new value used to replace or modify existing data

[†] The change set probability π_k is taken from this column of the first row for change set k.

to loads (fixed, dispatchable, real and/or reactive) than directly including individual change specifications for each of the corresponding entries in the bus and gen matrices. The row indices for load changes refer to bus numbers.

In addition to the normal direct modifications for generator cost parameters, there is also the option to scale or shift an entire cost function, either vertically or horizontally. This is often more convenient than manipulating the individual cost parameters directly, especially when dealing with a mix of polynomial and piecewise linear generator costs.

Table 9-2: Values for CT_TABLE Column

name	value	description
CT_TBUS	1	bus table
CT_TGEN	2	gen table
CT_TBRCH	3	branch table
CT_TAREABUS	4	area-wide change in bus table
$CT_TAREAGEN$	5	area-wide change in gen table
$CT_TAREABRCH$	6	area-wide change in branch table
CT_TLOAD	7	per bus load change [†]
CT_TAREALOAD	8	area-wide load change [†]
$CT_TGENCOST$	9	gencost table
CT_TAREAGENCOST	10	area-wide change in gencost table

 $^{^\}dagger$ Preferred method of modifying load, as opposed to manipulating $\tt bus$ and $\tt gen$ tables directly.

Normally, the CT_COL column contains the column index of the entry or entries in the data tables to be modified. And the CT_CHGTYPE and CT_NEWVAL columns specify, respectively, the type of change (replacement, scaling or adding) and the correspond-

ing replacement value, scale factor or constant to add, as shown in Table 9-3.

Table 9-3: Values for CT_CHGTYPE Column

name	value	description
CT_REP	1	replace old value by new one in CT_NEWVAL column
CT_REL	2	scale old value by factor in CT_NEWVAL column
$\mathtt{CT_ADD}$	3	add value in ${\tt CT_NEWVAL}$ column to old value

For load changes, the CT_COL column is not a column index, but rather a code that defines which loads at the specified bus(es) are to be modified, with the ability to select fixed loads only, dispatchable loads only or both, and for each whether or not to include the reactive load in the change. Similarly, the CT_COL column for generator cost modifications can be set to a special code to indicate a scaling or shifting of the entire corresponding cost function(s). The various options for the CT_COL column are summarized in Table 9-4.

Table 9-4: Values for CT_COL Column

name	value	description	
for CT_TABLE colum	$for \ \mathtt{CT_TABLE} \ column = \mathtt{CT_TLOAD} \ or \ \mathtt{CT_TAREALOAD}$		
CT_LOAD_ALL_PQ	1	modify all (fixed & dispatchable) loads, active and reactive	
CT_LOAD_FIX_PQ	2	modify fixed loads, active and reactive	
CT_LOAD_DIS_PQ	3	modify dispatchable loads, active and reactive	
CT_LOAD_ALL_P	4	modify all (fixed & dispatchable) loads, active power only	
CT_LOAD_FIX_P	5	modify fixed loads, active power only	
CT_LOAD_DIS_P	6	modify dispatchable loads, active power only	
for CT_TABLE colum	$n = CT_{-}$	TGENCOST or CT_TAREAGENCOST	
CT_MODCOST_F	-1	scales or shifts the cost function vertically †	
$\mathtt{CT_MODCOST_X}$	-2	scales or shifts the cost function horizontally †	
$otherwise^{\ddagger}$			
	n	index of column in data matrix to be modified	

 $^{^{\}dagger}$ Use CT_CHGTYPE column = CT_REL to scale the cost and CT_ADD to shift the cost.

[‡] Can also be used for CT_TGENCOST or CT_TAREAGENCOST in addition to the special codes above.

For example, setting up a chgtab matrix for the following four scenarios could be done as shown below.

- 1. Turn off generator 2 (10% probability).
- 2. Reduce the line rating of all lines to 95% of their nominal values (0.2% probability).
- 3. Scale all loads in area 2 (real & reactive, fixed & dispatchable) up by 10% (0.1% probability).
- 4. Decrease capacity of generator 3 and shift its cost function to the left both by 10 MW (5% probability).

```
chgtab = [ ...
                                2 GEN_STATUS
           0.1
                 CT_TGEN
                                                  CT_REP
                                                            0;
        1
            0.002 CT_TBRCH
                                O RATE_A
                                                  CT_REL
                                                            0.95;
            O.OO1 CT_TAREALOAD 2 CT_LOAD_ALL_PQ CT_REL
                                                            1.1;
            0.05 CT_TGEN
                                3 PMAX
                                                  CT_ADD
                                                         -10;
                                3 CT_MODCOST_X
            0.05
                 CT_TGENCOST
                                                  CT_ADD
                                                         -10;
];
```

A change table can be used to easily create modified cases from an existing base case with the apply_changes function. Given the chgtab from the example above, a new case with all lines derated by 5% can easily be created from an existing case mpc with the following line of code.

```
mpc_new = apply_changes(2, mpc, chgtab);
```

9.3.6 savechgtab

```
savechgtab(fname, chgtab)
savechgtab(fname, chgtab, warnings)
fname = savechgtab(fname, ...)
```

This function can be used to save a *change table* matrix, chgtab, to a file specified by fname. If the fname string ends with '.mat' it saves chgtab and warnings to a MAT-file as the variables chgtab and warnings, respectively. Otherwise, it saves an M-file function that returns the chgtab, with the optional warnings included in the comments, where warnings is a cell array of warning messages such as those returned by pssecon2chgtab.

9.4 Conversion between External and Internal Numbering

9.4.1 ext2int, int2ext

```
mpc_int = ext2int(mpc_ext)
mpc_int = ext2int(mpc_ext, mpopt)
mpc_ext = int2ext(mpc_int)
mpc_ext = int2ext(mpc_int, mpopt)
```

These functions convert a MATPOWER case struct from external to internal, and from internal to external numbering, respectively. ext2int first removes all isolated buses, off-line generators and branches, and any generators or branches connected to isolated buses. Then the buses are renumbered consecutively, beginning at 1.⁵⁶ Any 'ext2int' callback routines registered in the case are also invoked automatically. All of the related indexing information and the original data matrices are stored in an 'order' field in the struct to be used later by int2ext to perform the reverse conversions. If the case is already using internal numbering it is returned unchanged. The optional MATPOWER options struct (mpopt) input argument is only needed in conjunction with callback routines that depend on it, e.g. when toggle_softlims is on.

9.4.2 e2i_data, i2e_data

```
val = e2i_data(mpc, val, ordering)
val = e2i_data(mpc, val, ordering, dim)
val = i2e_data(mpc, val, oldval, ordering)
val = i2e_data(mpc, val, oldval, ordering, dim)
```

These functions can be used to convert other data structures from external to internal indexing and vice versa. When given a case struct (mpc) that has already been converted to internal indexing, $e2i_data$ can be used to convert other data structures as well by passing in 2 or 3 extra parameters in addition to the case struct. If the value passed in the second argument (val) is a column vector or cell array, it will be converted according to the ordering specified by the third argument (described below). If val is an n-dimensional matrix or cell array, then the optional fourth argument (dim, default = 1) can be used to specify which dimension to reorder. The return value in this case is the value passed in, converted to internal indexing.

⁵⁶In Matpower versions 4 through 7.0b1, ext2int also sorted the generators by increasing internal bus number, but beginning with version 7.0, the order of on-line generators is left unmodified.

The third argument, ordering, is used to indicate whether the data corresponds to bus-, gen- or branch-ordered data. It can be one of the following three strings: 'bus', 'gen' or 'branch'. For data structures with multiple blocks of data, ordered by bus, gen or branch, they can be converted with a single call by specifying ordering as a cell array of strings.

Any extra elements, rows, columns, etc. beyond those indicated in ordering, are not disturbed.

The function i2e_data performs the opposite conversion, from internal back to external indexing. It also assumes that mpc is using internal indexing, and the only difference is that it also includes an oldval argument used to initialize the return value before converting val to external indexing. In particular, any data corresponding to off-line gens or branches or isolated buses or any connected gens or branches will be taken from oldval, with val supplying the rest of the returned data.

9.4.3 e2i_field, i2e_field

```
mpc = e2i_field(mpc, field, ordering)
mpc = e2i_field(mpc, field, ordering, dim)
mpc = i2e_field(mpc, field, ordering)
mpc = i2e_field(mpc, field, ordering, dim)
```

These functions can be used to convert additional fields in mpc from external to internal indexing and vice versa. When given a case struct that has already been converted to internal indexing, e2i_field can be used to convert other fields as well by passing in 2 or 3 extra parameters in addition to the case struct.

The second argument (field) is a string or cell array of strings, specifying a field in the case struct whose value should be converted by a corresponding call to e2i_data. The field can contain either a numeric or a cell array. The converted value is stored back in the specified field, the original value is saved for later use and the updated case struct is returned. If field is a cell array of strings, they specify nested fields.

The third and optional fourth arguments (ordering and dim) are simply passed along to the call to e2i_data.

Similarly, i2e_field performs the opposite conversion, from internal back to external indexing. It also assumes that mpc is using internal indexing and utilizes the original data stored by e2i_field, calling i2e_data to do the conversion work.

9.5 Forming Standard Power Systems Matrices

9.5.1 makeB

```
[Bp, Bpp] = makeB(mpc, alg)
[Bp, Bpp] = makeB(baseMVA, bus, branch, alg)
```

The makeB function builds the two matrices B' and B'' used in the fast-decoupled power flow. The alg, which can take values 'FDXB' or 'FDBX', determines whether the matrices returned correspond to the XB or BX version of the fast-decoupled power flow. Bus numbers must be consecutive beginning at 1 (i.e. internal ordering).

9.5.2 makeBdc

```
[Bbus, Bf, Pbusinj, Pfinj] = makeBdc(mpc)
[Bbus, Bf, Pbusinj, Pfinj] = makeBdc(baseMVA, bus, branch)
```

The function builds the B matrices, B_{dc} (Bbus) and B_f (Bf), and phase shift injections, P_{dc} (Pbusinj) and $P_{f,\text{shift}}$ (Pfinj), for the DC power flow model as described in (3.29) and (4.15). Bus numbers must be consecutive beginning at 1 (i.e. internal ordering).

9.5.3 makeJac

```
J = makeJac(mpc)
J = makeJac(mpc, fullJac)
[J, Ybus, Yf, Yt] = makejac(mpc)
```

The makeJac function forms the power flow Jacobian and, optionally, the system admittance matrices. Bus numbers in the input case must be consecutive beginning at 1 (i.e. internal indexing). If the fullJac argument is present and true, it returns the full Jacobian (sensitivities of all bus injections with respect to all voltage angles and magnitudes) as opposed to the reduced version used in the Newton power flow updates.

9.5.4 makeLODF

```
PTDF = makePTDF(mpc)
LODF = makeLODF(mpc.branch, PTDF)
```

The makeLODF function forms the DC line outage distribution factor matrix for a given PTDF. The matrix is $n_{br} \times n_{br}$, where n_{br} is the number of branches. See Section 4.5 on Linear Shift Factors for more details.

9.5.5 makePTDF

```
H = makePTDF(mpc)
H = makePTDF(mpc, slack)
H = makePTDF(baseMVA, bus, branch)
H = makePTDF(baseMVA, bus, branch, slack)
```

The makePTDF function returns the DC PTDF matrix for a given choice of slack. The matrix is $n_{br} \times n_b$, where n_{br} is the number of branches and n_b is the number of buses. The slack can be a scalar (single slack bus) or an $n_b \times 1$ column vector of weights specifying the proportion of the slack taken up at each bus. If the slack is not specified, the reference bus is used by default. Bus numbers must be consecutive beginning at 1 (i.e. internal ordering). See Section 4.5 on Linear Shift Factors for more details.

9.5.6 makeYbus

```
[Ybus, Yf, Yt] = makeYbus(mpc)
[Ybus, Yf, Yt] = makeYbus(baseMVA, bus, branch)
```

The makeYbus function builds the bus admittance matrix and branch admittance matrices from (3.11)-(3.13). Bus numbers in the input case must be consecutive beginning at 1 (i.e. internal indexing).

9.6 Miscellaneous

9.6.1 define_constants

```
define_constants
```

The define_constants script is a convenience script that defines a set of useful constants, mostly to be used as named column indices into the bus, branch, gen and gencost data matrices. The purpose is to avoid having to remember column numbers and to allow code to be more robust against potential future changes to the MATPOWER case data format. It also defines constants for the change tables used by apply_changes.

Specifically, it includes all of the constants defined by idx_bus, idx_brch, idx_gen, idx_cost and idx_ct.

9.6.2 feval_w_path

```
[y1, ..., yn] = feval_w_path(fpath, f, x1, ..., xn)
```

The feval_w_path function is identical to MATLAB's own feval, except that the function f need not be in the MATLAB path if it is defined in a file in the path specified by fpath. Assumes that the current working directory is always first in the MATLAB path.

9.6.3 have_fcn

```
TorF = have_fcn(tag)
TorF = have_fcn(tag, toggle)
ver_str = have_fcn(tag, 'vstr')
ver_num = have_fcn(tag, 'vnum')
rdate = have_fcn(tag, 'date')
info = have_fcn(tag, 'all')
```

The have_fcn function provides a unified mechanism for testing for optional functionality, such as the presence of certain solvers, or to detect whether the code is running under MATLAB or Octave. Since its results are cached they allow for a very quick way to check frequently for functionality that may initially be a bit more costly to determine. For installed functionality, have_fcn also determines the installed version and release date, if possible. The optional second argument, when it is a string, defines which value is returned, as follows:

- empty 1 if optional functionality is available, 0 if not available
- 'vstr' version number as a string (e.g. '3.11.4')
- 'vnum' version number as numeric value (e.g. 3.011004)
- 'date' release date as a string (e.g. '20-Jan-2015')
- 'all' struct with fields named av (for "availability"), vstr, vnum and date, and values corresponding to each of the above, respectively.

Alternatively, the optional functionality specified by tag can be toggled OFF or ON by calling have_fcn with a numeric second argument toggle with one of the following values:

- 0 turn OFF the optional functionality
- 1 turn ON the optional functionality (if available)
- -1 toggle the ON/OFF state of the optional functionality

9.6.4 mpopt2qpopt

```
qpopt = mpopt2qpopt(mpopt)
qpopt = mpopt2qpopt(mpopt, model)
qpopt = mpopt2qpopt(mpopt, model, alg)
```

The mpopt2qpopt function returns an options struct suitable for qps_matpower, miqps_matpower or one of the solver specific equivalents. It is constructed from the relevant portions of mpopt, a MATPOWER options struct. The model argument specifies whether the problem to be solved is an LP, QP, MILP or MIQP problem to allow for the selection of a suitable default solver. The final alg argument allows the solver to be set explicitly (in qpopt.alg). By default this value is taken from mpopt.opf.dc.solver.

When the solver is set to 'DEFAULT', this function also selects the best available solver that is applicable⁵⁷ to the specific problem class, based on the following precedence: Gurobi, CPLEX, MOSEK, Optimization Toolbox, GLPK, BPMPD, MIPS.

⁵⁷GLPK is not available for problems with quadratic costs (QP and MIQP), BPMPD and MIPS are not available for mixed integer problems (MILP and MIQP), and the Optimization Toolbox is not an option for problems that combine the two (MIQP).

9.6.5 mpver

```
mpver
v = mpver
v = mpver('all')
```

The mpver function returns the current MATPOWER version number. With the optional 'all' argument, it returns a struct with the fields 'Name', 'Version', 'Release' and 'Date' (all strings). Calling mpver without assigning the return value prints the version and release date of the current installation of MATPOWER, MATLAB (or Octave), the Optimization Toolbox, MIPS and any optional MATPOWER packages.

9.6.6 nested_struct_copy

```
ds = nested_struct_copy(d, s)
ds = nested_struct_copy(d, s, opt)
```

The nested_struct_copy function copies values from a source struct s to a destination struct d in a nested, recursive manner. That is, the value of each field in s is copied directly to the corresponding field in d, unless that value is itself a struct, in which case the copy is done via a recursive call to nested_struct_copy. Certain aspects of the copy behavior can be controlled via the optional options struct opt, including the possible checking of valid field names.

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Appendix A MIPS – MATPOWER Interior Point Solver

Beginning with version 4, MATPOWER includes a new primal-dual interior point solver called MIPS, for MATPOWER Interior Point Solver. It is implemented in pure-MATLAB code, derived from the MEX implementation of the algorithms described in [8,43].

This solver has application outside of MATPOWER to general nonlinear optimization problems of the following form:

$$\min_{x} f(x) \tag{A.1}$$

subject to

$$g(x) = 0 (A.2)$$

$$h(x) \le 0 \tag{A.3}$$

$$l \le Ax \le u \tag{A.4}$$

$$x_{\min} \le x \le x_{\max}$$
 (A.5)

where $f: \mathbb{R}^n \to \mathbb{R}$, $g: \mathbb{R}^n \to \mathbb{R}^m$ and $h: \mathbb{R}^n \to \mathbb{R}^p$.

The solver is implemented by the mips function, which can be called as follows,

```
[x, f, exitflag, output, lambda] = ...
mips(f_fcn, x0, A, l, u, xmin, xmax, gh_fcn, hess_fcn, opt);
```

where the input and output arguments are described in Tables A-1 and A-2, respectively. Alternatively, the input arguments can be packaged as fields in a problem struct and passed in as a single argument, where all fields except f_fcn and x0 are optional.

```
[x, f, exitflag, output, lambda] = mips(problem);
```

The calling syntax is nearly identical to that used by fmincon from MATLAB's Optimization Toolbox. The primary difference is that the linear constraints are specified in terms of a single doubly-bounded linear function ($l \le Ax \le u$) as opposed to separate equality constrained ($A_{eq}x = b_{eq}$) and upper bounded ($Ax \le b$) functions. Internally, equality constraints are handled explicitly and determined at run-time based on the values of l and u.

The user-defined functions for evaluating the objective function, constraints and Hessian are identical to those required by fmincon, with one exception described

Table A-1: Input Arguments for mips[†]

name	description
f_fcn	Handle to a function that evaluates the objective function, its gradients and Hessian [‡] for a given value of x . Calling syntax for this function: [f, df, d2f] = f_fcn(x)
x0	Starting value of optimization vector x .
A, 1, u	Define the optional linear constraints $l \leq Ax \leq u$. Default values for the elements of 1 and u are -Inf and Inf, respectively.
xmin, xmax	Optional lower and upper bounds on the x variables, defaults are -Inf and Inf, respectively.
gh_fcn	Handle to function that evaluates the optional nonlinear constraints and their gradients for a given value of x . Calling syntax for this function is: $[h, g, dh, dg] = gh_f cn(x)$
hess_fcn	Handle to function that computes the Hessian [‡] of the Lagrangian for given values of x , λ and μ , where λ and μ are the multipliers on the equality and inequality constraints, g and h , respectively. The calling syntax for this function is: Lxx = hess_fcn(x, lam, cost_mult), where λ = lam.eqnonlin, μ = lam.ineqnonlin and cost_mult is a parameter used
	to scale the objective function
opt	Optional options structure with fields, all of which are also optional, described in Table A-3.
problem	Alternative, single argument input struct with fields corresponding to arguments above.

[†] All inputs are optional except f_fcn and x0.

below for the Hessian evaluation function. Specifically, f_fcn should return f as the scalar objective function value f(x), df as an $n \times 1$ vector equal to ∇f and, unless gh_fcn is provided and the Hessian is computed by hess_fcn, d2f as an $n \times n$ matrix equal to the Hessian $\frac{\partial^2 f}{\partial x^2}$. Similarly, the constraint evaluation function gh_fcn must return the $m \times 1$ vector of nonlinear equality constraint violations g(x), the $p \times 1$ vector of nonlinear inequality constraint violations h(x) along with their gradients in dg and dh. Here dg is an $n \times m$ matrix whose j^{th} column is ∇g_j and dh is $n \times p$, with j^{th} column equal to ∇h_j . Finally, for cases with nonlinear constraints, hess_fcn returns the $n \times n$ Hessian $\frac{\partial^2 \mathcal{L}}{\partial x^2}$ of the Lagrangian function

$$\mathcal{L}(x,\lambda,\mu,\sigma) = \sigma f(x) + \lambda^{\mathsf{T}} g(x) + \mu^{\mathsf{T}} h(x)$$
(A.6)

for given values of the multipliers λ and μ , where σ is the cost_mult scale factor for the objective function. Unlike fmincon, mips passes this scale factor to the Hessian evaluation function in the 3rd argument.

[‡] If gh_fcn is provided then hess_fcn is also required. Specifically, if there are nonlinear constraints, the Hessian information must provided by the hess_fcn function and it need not be computed in f_fcn.

Table A-2: Output Arguments for mips

name	description			
х	solution vector	solution vector		
f	final objective	function value		
exitflag	exit flag			
	1 - first orde	er optimality conditions satisfied		
	0 - maximur	n number of iterations reached		
	-1 – numerica	ally failed		
output	output struct with fields			
	iterations	number of iterations performed		
	hist	struct array with trajectories of the following: feascond,		
		gradcond, compcond, costcond, gamma, stepsize, obj, alphap,		
		alphad		
	message	exit message		
lambda	struct contain	ing the Langrange and Kuhn-Tucker multipliers on the con-		
	straints, with f			
	eqnonlin	nonlinear equality constraints		
	ineqnonlin	nonlinear inequality constraints		
	$\mathtt{mu_l}$	lower (left-hand) limit on linear constraints		
	mu_u	upper (right-hand) limit on linear constraints		
	lower	lower bound on optimization variables		
	upper	upper bound on optimization variables		

The use of nargout in f_fcn and gh_fcn is recommended so that the gradients and Hessian are only computed when required.

A.1 Example 1

The following code shows a simple example of using mips to solve a 2-dimensional unconstrained optimization of Rosenbrock's "banana" function⁵⁸

$$f(x) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2. (A.7)$$

First, create a MATLAB function that will evaluate the objective function, its gradients and Hessian, for a given value of x. In this case, the coefficient of the first term is defined as a paramter a.

⁵⁸https://en.wikipedia.org/wiki/Rosenbrock_function

Table A-3: Options for mips[†]

name	default	description
opt.verbose	0	controls level of progress output displayed
		0 – print no progress info
		1 – print a little progress info
		2 – print a lot of progress info
		3 – print all progress info
opt.feastol	10^{-6}	termination tolerance for feasibility condition
opt.gradtol	10^{-6}	termination tolerance for gradient condition
opt.comptol	10^{-6}	termination tolerance for complementarity condition
opt.costtol	10^{-6}	termination tolerance for cost condition
opt.max_it	150	maximum number of iterations
opt.step_control	0	set to 1 to enable step-size control
${\tt opt.sc.red_it}$	20	max number of step-size reductions if step-control is on
opt.cost_mult	1	cost multiplier used to scale the objective function for improved
		conditioning. Note: This value is also passed as the 3 rd argu-
		ment to the Hessian evaluation function so that it can appro-
		priately scale the objective function term in the Hessian of the
		Lagrangian.
opt.xi	0.99995	ξ constant used in α updates in (A.46) and (A.47)
opt.sigma	0.1	centering parameter σ used in γ update in (A.52)
opt.z0	1	used to initialize elements of slack variable Z
${\tt opt.alpha_min}$	10^{-8}	algorithm returns "Numerically Failed" if the α_p or α_d from
		(A.46) and $(A.47)$ become smaller than this value
${\tt opt.rho_min}$	0.95	lower bound on ρ_t corresponding to $1 - \eta$ in Fig. 5 in [8]
${\tt opt.rho_max}$	1.05	upper bound on ρ_t corresponding to $1 + \eta$ in Fig. 5 in [8]
${\tt opt.mu_threshold}$	10^{-5}	Kuhn-Tucker multipliers smaller than this value for non-binding
	10	constraints are forced to zero
${\tt opt.max_stepsize}$	10^{10}	algorithm returns "Numerically Failed" if the 2-norm of the New-
		ton step $\begin{vmatrix} \Delta X \\ \Delta \lambda \end{vmatrix}$ from (A.45) exceeds this value

Then, create a handle to the function, defining the value of the paramter a to be 100, set up the starting value of x, and call the mips function to solve it.

```
>> f_fcn = @(x)banana(x, 100);
>> x0 = [-1.9; 2];
>> [x, f] = mips(f_fcn, x0)
     1
     0
```

Example 2 **A.2**

The second example⁵⁹ solves the following 3-dimensional constrained optimization, printing the details of the solver's progress:

$$\min_{x} f(x) = -x_1 x_2 - x_2 x_3 \tag{A.8}$$

subject to

$$x_1^2 - x_2^2 + x_3^2 - 2 \le 0$$
 (A.9)
 $x_1^2 + x_2^2 + x_3^2 - 10 \le 0$. (A.10)

$$x_1^2 + x_2^2 + x_3^2 - 10 \le 0.$$
 (A.10)

First, create a MATLAB function to evaluate the objective function and its gradients,60

⁵⁹From https://en.wikipedia.org/wiki/Nonlinear_programming#3-dimensional_example.

⁶⁰Since the problem has nonlinear constraints and the Hessian is provided by hess_fcn, this function will never be called with three output arguments, so the code to compute d2f is actually not necessary.

one to evaluate the constraints, in this case inequalities only, and their gradients,

```
function [h, g, dh, dg] = gh2(x)

h = [ 1 -1 1; 1 1 1] * x.^2 + [-2; -10];

dh = 2 * [x(1) x(1); -x(2) x(2); x(3) x(3)];

g = []; dg = [];
```

and another to evaluate the Hessian of the Lagrangian.

Then create a problem struct with handles to these functions, a starting value for x and an option to print the solver's progress. Finally, pass this struct to mips to solve the problem and print some of the return values to get the output below.

```
function example2
problem = struct( ...
    'f_fcn',
                0(x)f2(x), ...
                0(x)gh2(x), ...
    'gh_fcn',
    'hess_fcn', @(x, lam, cost_mult)hess2(x, lam, cost_mult), ...
              [1; 1; 0], ...
    'opt',
               struct('verbose', 2) ...
[x, f, exitflag, output, lambda] = mips(problem);
fprintf('\nf = %g exitflag = %d\n', f, exitflag);
fprintf('\nx = \n');
fprintf('
           %g\n', x);
fprintf('\nlambda.ineqnonlin =\n');
fprintf('
            %g\n', lambda.ineqnonlin);
```

>> example2 MATPOWER Interior Point Solver MIPS, Version 1.3.1, 20-Jun-2019						
(using built-in linear solver)						
it	objective	step size	feascond	gradcond	compcond	costcond
0	-1		0		5	0
	-5.3250167		-	0.894235		
2	-7.4708991	0.97413	0.129183	0.00936418	0.117278	0.339269
3	-7.0553031	0.10406	0	0.00174933	0.0196518	0.0490616
4	-7.0686267	0.034574	0	0.00041301	0.0030084	0.00165402
5	-7.0706104	0.0065191	0	1.53531e-05	0.000337971	0.000245844
6	-7.0710134	0.00062152	0	1.22094e-07	3.41308e-05	4.99387e-05
7	-7.0710623	5.7217e-05	0	9.84879e-10	3.41587e-06	6.05875e-06
8	-7.0710673	5.6761e-06	0	9.73527e-12	3.41615e-07	6.15483e-07
Converged!						
f = -7.07107 exitflag = 1						
x =						
1.58114						
2.23607						
1.	58114					
lambda.ineqnonlin =						
0						
0.	707107					

More example problems for mips can be found in t_mips.m.

A.3 Quadratic Programming Solver

A convenience wrapper function called qps_mips is provided to make it trivial to set up and solve linear programming (LP) and quadratic programming (QP) problems of the following form:

$$\min_{x} \frac{1}{2} x^{\mathsf{T}} H x + c^{\mathsf{T}} x \tag{A.11}$$

subject to

$$l \le Ax \le u \tag{A.12}$$

$$x_{\min} \le x \le x_{\max}. \tag{A.13}$$

Instead of a function handle, the objective function is specified in terms of the paramters H and c of quadratic cost coefficients. Internally, qps_mips passes mips

the handle of a function that uses these paramters to evaluate the objective function, gradients and Hessian.

The calling syntax for qps_mips is similar to that used by quadprog from the MATLAB Optimization Toolbox.

```
[x, f, exitflag, output, lambda] = qps_mips(H, c, A, 1, u, xmin, xmax, x0, opt);
```

Alternatively, the input arguments can be packaged as fields in a problem struct and passed in as a single argument, where all fields except H, c, A and 1 are optional.

```
[x, f, exitflag, output, lambda] = qps_mips(problem);
```

Aside from H and c, all input and output arguments correspond exactly to the same arguments for mips as described in Tables A-1 and A-2.

As with mips and fmincon, the primary difference between the calling syntax for qps_mips and quadprog is that the linear constraints are specified in terms of a single doubly-bounded linear function $(l \le Ax \le u)$ as opposed to separate equality constrained $(A_{eq}x = b_{eq})$ and upper bounded $(Ax \le b)$ functions.

MATPOWER also includes another wrapper function qps_matpower that provides a consistent interface for all of the QP and LP solvers it has available. This interface is identical to that used by qps_mips with the exception of the structure of the opt input argument. The solver is chosen according to the value of opt.alg. See the help for qps_matpower for details.

Several examples of using qps_matpower to solve LP and QP problems can be found in t_qps_matpower.m.

A.4 Primal-Dual Interior Point Algorithm

This section provides some details on the primal-dual interior point algorithm used by MIPS and described in [8,43].

A.4.1 Notation

For a scalar function $f: \mathbb{R}^n \to \mathbb{R}$ of a real vector $X = \begin{bmatrix} x_1 & x_2 & \cdots & x_n \end{bmatrix}^\mathsf{T}$, we use the following notation for the first derivatives (transpose of the gradient):

$$f_X = \frac{\partial f}{\partial X} = \begin{bmatrix} \frac{\partial f}{\partial x_1} & \frac{\partial f}{\partial x_2} & \cdots & \frac{\partial f}{\partial x_n} \end{bmatrix}.$$
 (A.14)

The matrix of second partial derivatives, the Hessian of f, is:

$$f_{XX} = \frac{\partial^2 f}{\partial X^2} = \frac{\partial}{\partial X} \left(\frac{\partial f}{\partial X} \right)^\mathsf{T} = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \cdots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \cdots & \frac{\partial^2 f}{\partial x_n^2} \end{bmatrix}. \tag{A.15}$$

For a vector function $F: \mathbb{R}^n \to \mathbb{R}^m$ of a vector X, where

$$F(X) = \begin{bmatrix} f_1(X) & f_2(X) & \cdots & f_m(X) \end{bmatrix}^\mathsf{T}$$
 (A.16)

the first derivatives form the Jacobian matrix, where row i is the transpose of the gradient of f_i

$$F_X = \frac{\partial F}{\partial X} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \cdots & \frac{\partial f_m}{\partial x_n} \end{bmatrix}. \tag{A.17}$$

In these derivations, the full 3-dimensional set of second partial derivatives of F will not be computed. Instead a matrix of partial derivatives will be formed by computing the Jacobian of the vector function obtained by multiplying the transpose of the Jacobian of F by a vector λ , using the following notation

$$F_{XX}(\lambda) = \frac{\partial}{\partial X} \left(F_X^{\mathsf{T}} \lambda \right). \tag{A.18}$$

Please note also that [A] is used to denote a diagonal matrix with vector A on the diagonal and e is a vector of all ones.

A.4.2 Problem Formulation and Lagrangian

The primal-dual interior point method used by MIPS solves a problem of the form:

$$\min_{X} f(X) \tag{A.19}$$

subject to

$$G(X) = 0 (A.20)$$

$$H(X) \le 0 \tag{A.21}$$

where the linear constraints and variable bounds from (A.4) and (A.5) have been incorporated into G(X) and H(X). The approach taken involves converting the n_i

inequality constraints into equality constraints using a barrier function and vector of positive slack variables Z.

$$\min_{X} \left[f(X) - \gamma \sum_{m=1}^{n_i} \ln(Z_m) \right] \tag{A.22}$$

subject to

$$G(X) = 0 (A.23)$$

$$H(X) + Z = 0 \tag{A.24}$$

$$Z > 0 \tag{A.25}$$

As the parameter of perturbation γ approaches zero, the solution to this problem approaches that of the original problem.

For a given value of γ , the Lagrangian for this equality constrained problem is

$$\mathcal{L}^{\gamma}(X, Z, \lambda, \mu) = f(X) + \lambda^{\mathsf{T}} G(X) + \mu^{\mathsf{T}} (H(X) + Z) - \gamma \sum_{m=1}^{n_i} \ln(Z_m). \tag{A.26}$$

Taking the partial derivatives with respect to each of the variables yields:

$$\mathcal{L}_{X}^{\gamma}(X, Z, \lambda, \mu) = f_{X} + \lambda^{\mathsf{T}} G_{X} + \mu^{\mathsf{T}} H_{X}$$
 (A.27)

$$\mathcal{L}_{Z}^{\gamma}(X, Z, \lambda, \mu) = \mu^{\mathsf{T}} - \gamma e^{\mathsf{T}} [Z]^{-1}$$
(A.28)

$$\mathcal{L}_{\lambda}^{\gamma}(X, Z, \lambda, \mu) = G^{\mathsf{T}}(X) \tag{A.29}$$

$$\mathcal{L}_{n}^{\gamma}(X, Z, \lambda, \mu) = H^{\mathsf{T}}(X) + Z^{\mathsf{T}}. \tag{A.30}$$

And the Hessian of the Lagrangian with respect to X is given by

$$\mathcal{L}_{XX}^{\gamma}(X, Z, \lambda, \mu) = f_{XX} + G_{XX}(\lambda) + H_{XX}(\mu). \tag{A.31}$$

A.4.3 First Order Optimality Conditions

The first order optimality (Karush-Kuhn-Tucker) conditions for this problem are satisfied when the partial derivatives of the Lagrangian above are all set to zero:

$$F(X, Z, \lambda, \mu) = 0 \tag{A.32}$$

$$Z > 0 \tag{A.33}$$

$$\mu > 0 \tag{A.34}$$

where

$$F(X, Z, \lambda, \mu) = \begin{bmatrix} \mathcal{L}_X^{\gamma \mathsf{T}} \\ [\mu] Z - \gamma e \\ G(X) \\ H(X) + Z \end{bmatrix} = \begin{bmatrix} f_X^{\mathsf{T}} + G_X^{\mathsf{T}} \lambda + H_X^{\mathsf{T}} \mu \\ [\mu] Z - \gamma e \\ G(X) \\ H(X) + Z \end{bmatrix}. \tag{A.35}$$

A.4.4 Newton Step

The first order optimality conditions are solved using Newton's method. The Newton update step can be written as follows:

$$\begin{bmatrix} F_X & F_Z & F_{\lambda} & F_{\mu} \end{bmatrix} \begin{bmatrix} \Delta X \\ \Delta Z \\ \Delta \lambda \\ \Delta \mu \end{bmatrix} = -F(X, Z, \lambda, \mu)$$
 (A.36)

$$\begin{bmatrix} \mathcal{L}_{XX}^{\gamma} & 0 & G_{X}^{\mathsf{T}} & H_{X}^{\mathsf{T}} \\ 0 & [\mu] & 0 & [Z] \\ G_{X} & 0 & 0 & 0 \\ H_{X} & I & 0 & 0 \end{bmatrix} \begin{bmatrix} \Delta X \\ \Delta Z \\ \Delta \lambda \\ \Delta \mu \end{bmatrix} = - \begin{bmatrix} \mathcal{L}_{X}^{\gamma \mathsf{T}} \\ [\mu] Z - \gamma e \\ G(X) \\ H(X) + Z \end{bmatrix}. \tag{A.37}$$

This set of equations can be simplified and reduced to a smaller set of equations by solving explicitly for $\Delta\mu$ in terms of ΔZ and for ΔZ in terms of ΔX . Taking the $2^{\rm nd}$ row of (A.37) and solving for $\Delta\mu$ we get

$$[\mu] \Delta Z + [Z] \Delta \mu = -[\mu] Z + \gamma e$$

$$[Z] \Delta \mu = -[Z] \mu + \gamma e - [\mu] \Delta Z$$

$$\Delta \mu = -\mu + [Z]^{-1} (\gamma e - [\mu] \Delta Z). \tag{A.38}$$

Solving the 4th row of (A.37) for ΔZ yields

$$H_X \Delta X + \Delta Z = -H(X) - Z$$

$$\Delta Z = -H(X) - Z - H_X \Delta X. \tag{A.39}$$

Then, substituting (A.38) and (A.39) into the 1st row of (A.37) results in

$$\mathcal{L}_{XX}^{\gamma} \Delta X + G_{X}^{\mathsf{T}} \Delta \lambda + H_{X}^{\mathsf{T}} \Delta \mu = -\mathcal{L}_{X}^{\gamma\mathsf{T}} \\
\mathcal{L}_{XX}^{\gamma} \Delta X + G_{X}^{\mathsf{T}} \Delta \lambda + H_{X}^{\mathsf{T}} (-\mu + [Z]^{-1} (\gamma e - [\mu] \Delta Z)) = -\mathcal{L}_{X}^{\gamma\mathsf{T}} \\
\mathcal{L}_{XX}^{\gamma} \Delta X + G_{X}^{\mathsf{T}} \Delta \lambda \\
+ H_{X}^{\mathsf{T}} (-\mu + [Z]^{-1} (\gamma e - [\mu] (-H(X) - Z - H_{X} \Delta X))) = -\mathcal{L}_{X}^{\gamma\mathsf{T}} \\
\mathcal{L}_{XX}^{\gamma} \Delta X + G_{X}^{\mathsf{T}} \Delta \lambda - H_{X}^{\mathsf{T}} \mu + H_{X}^{\mathsf{T}} [Z]^{-1} \gamma e \\
+ H_{X}^{\mathsf{T}} [Z]^{-1} [\mu] H(X) + H_{X}^{\mathsf{T}} [Z]^{-1} [Z] \mu + H_{X}^{\mathsf{T}} [Z]^{-1} [\mu] H_{X} \Delta X = -\mathcal{L}_{X}^{\gamma\mathsf{T}} \\
(\mathcal{L}_{XX}^{\gamma} + H_{X}^{\mathsf{T}} [Z]^{-1} [\mu] H_{X}) \Delta X + G_{X}^{\mathsf{T}} \Delta \lambda \\
+ H_{X}^{\mathsf{T}} [Z]^{-1} (\gamma e + [\mu] H(X)) = -\mathcal{L}_{X}^{\gamma\mathsf{T}} \\
M \Delta X + G_{X}^{\mathsf{T}} \Delta \lambda = -N \quad (A.40)$$

where

$$M \equiv \mathcal{L}_{XX}^{\gamma} + H_X^{\mathsf{T}} [Z]^{-1} [\mu] H_X \tag{A.41}$$

$$= f_{XX} + G_{XX}(\lambda) + H_{XX}(\mu) + H_X^{\mathsf{T}}[Z]^{-1}[\mu] H_X \tag{A.42}$$

and

$$N \equiv \mathcal{L}_X^{\gamma \mathsf{T}} + H_X^{\mathsf{T}} [Z]^{-1} (\gamma e + [\mu] H(X))$$
(A.43)

$$= f_X^{\mathsf{T}} + G_X^{\mathsf{T}} \lambda + H_X^{\mathsf{T}} \mu + H_X^{\mathsf{T}} [Z]^{-1} (\gamma e + [\mu] H(X)). \tag{A.44}$$

Combining (A.40) and the $3^{\rm rd}$ row of (A.37) results in a system of equations of reduced size:

$$\begin{bmatrix} M & G_X^{\mathsf{T}} \\ G_X & 0 \end{bmatrix} \begin{bmatrix} \Delta X \\ \Delta \lambda \end{bmatrix} = \begin{bmatrix} -N \\ -G(X) \end{bmatrix}. \tag{A.45}$$

The Newton update can then be computed in the following 3 steps:

- 1. Compute ΔX and $\Delta \lambda$ from (A.45).
- 2. Compute ΔZ from (A.39).
- 3. Compute $\Delta \mu$ from (A.38).

In order to maintain strict feasibility of the trial solution, the algorithm truncates the Newton step by scaling the primal and dual variables by α_p and α_d , respectively,

where these scale factors are computed as follows:

$$\alpha_p = \min\left(\xi \min_{\Delta Z_m < 0} \left(-\frac{Z_m}{\Delta Z_m}\right), 1\right)$$
 (A.46)

$$\alpha_d = \min\left(\xi \min_{\Delta \mu_m < 0} \left(-\frac{\mu_m}{\Delta \mu_m} \right), 1 \right) \tag{A.47}$$

resulting in the variable updates below.

$$X \leftarrow X + \alpha_p \Delta X \tag{A.48}$$

$$Z \leftarrow Z + \alpha_p \Delta Z$$
 (A.49)

$$\lambda \leftarrow \lambda + \alpha_d \Delta \lambda \tag{A.50}$$

$$\mu \leftarrow \mu + \alpha_d \Delta \mu \tag{A.51}$$

The parameter ξ is a constant scalar with a value slightly less than one. In MIPS, ξ is set to 0.99995.

In this method, during the Newton-like iterations, the perturbation parameter γ must converge to zero in order to satisfy the first order optimality conditions of the original problem. MIPS uses the following rule to update γ at each iteration, after updating Z and μ :

$$\gamma \leftarrow \sigma \frac{Z^{\mathsf{T}} \mu}{n_i} \tag{A.52}$$

where σ is a scalar constant between 0 and 1. In MIPS, σ is set to 0.1.

Appendix B Data File Format

There are two versions of the MATPOWER case file format. MATPOWER versions 3.0.0 and earlier used the version 1 format internally. Subsequent versions of MATPOWER have used the version 2 format described below, though version 1 files are still handled, and converted automatically, by the loadcase and savecase functions.

In the version 2 format, the input data for MATPOWER are specified in a set of data matrices packaged as the fields of a MATLAB struct, referred to as a "MATPOWER case" struct and conventionally denoted by the variable mpc. This struct is typically defined in a case file, either a function M-file whose return value is the mpc struct or a MAT-file that defines a variable named mpc when loaded. The fields of this struct are baseMVA, bus, branch, gen and, optionally, gencost. The baseMVA field is a scalar and the rest are matrices. Each row in the data matrices corresponds to a single bus, branch, or generator and the columns are similar to the columns in the standard IEEE and PTI formats. The mpc struct also has a version field whose value is a string set to the current MATPOWER case version, currently '2' by default. The version 1 case format defines the data matrices as individual variables rather than fields of a struct, and some do not include all of the columns defined in version 2.

Numerous examples can be found in the case files listed in Table D-18 in Appendix D. The case files created by savecase use a tab-delimited format for the data matrices to make it simple to transfer data seamlessly back and forth between a text editor and a spreadsheet via simple copy and paste.

The details of the MATPOWER case format are given in the tables below and can also be accessed by typing help caseformat at the MATLAB prompt. First, the baseMVA field is a simple scalar value specifying the system MVA base used for converting power into per unit quantities. For convenience and code portability, idx_bus defines a set of constants to be used as named indices into the columns of the bus matrix. Similarly, idx_brch, idx_gen and idx_cost define names for the columns of branch, gen and gencost, respectively. The script define_constants provides a simple way to define all the usual constants at one shot. These are the names that appear in the first column of the tables below.

The MATPOWER case format also allows for additional fields to be included in the structure. The OPF is designed to recognize fields named A, 1, u, H, Cw, N, fparm, z0, z1 and zu as parameters used to directly extend the OPF formulation as described in Section 7.1. Additional standard optional fields include bus_name, gentype and genfuel. Other user-defined fields may also be included, such as the reserves field used in the example code throughout Section 7.3. The loadcase function will automatically load any extra fields from a case file and, if the appropriate 'savecase' callback function (see Section 7.3.5) is added via add_userfcn, savecase will also save them back to a case file.

Table B-1: Bus Data (mpc.bus)

name	column	description
BUS_I	1	bus number (positive integer)
BUS_TYPE	2	bus type $(1 = PQ, 2 = PV, 3 = ref, 4 = isolated)$
PD	3	real power demand (MW)
QD	4	reactive power demand (MVAr)
GS	5	shunt conductance (MW demanded at $V = 1.0$ p.u.)
BS	6	shunt susceptance (MVAr injected at $V = 1.0$ p.u.)
BUS_AREA	7	area number (positive integer)
VM	8	voltage magnitude (p.u.)
VA	9	voltage angle (degrees)
BASE_KV	10	base voltage (kV)
ZONE	11	loss zone (positive integer)
VMAX	12	maximum voltage magnitude (p.u.)
VMIN	13	minimum voltage magnitude (p.u.)
$\mathtt{LAM_P}^\dagger$	14	Lagrange multiplier on real power mismatch (u/MW)
$\mathtt{LAM}_{Q}^{\dagger}$	15	Lagrange multiplier on reactive power mismatch $(u/MVAr)$
$\mathtt{MU}_{-}\mathtt{VMAX}^{\dagger}$	16	Kuhn-Tucker multiplier on upper voltage limit $(u/p.u.)$
MU_VMIN†	17	Kuhn-Tucker multiplier on lower voltage limit $(u/p.u.)$

 $^{^\}dagger$ Included in OPF output, typically not included (or ignored) in input matrix. Here we assume the objective function has units u.

⁶¹All three of these are cell arrays of strings. See gentypes and genfuels for more information on the corresponding fields.

Table B-2: Generator Data (mpc.gen)

nomo	column	description
name	Corumn	description
GEN_BUS	1	bus number
PG	2	real power output (MW)
QG	3	reactive power output (MVAr)
QMAX	4	maximum reactive power output (MVAr)
QMIN	5	minimum reactive power output (MVAr)
VG^{\ddagger}	6	voltage magnitude setpoint (p.u.)
MBASE	7	total MVA base of machine, defaults to baseMVA
GEN_STATUS	8	machine status, > 0 = machine in-service
GLN_DIATOD	O	machine status, $\leq 0 = \text{machine out-of-service}$
PMAX	9	maximum real power output (MW)
PMIN	10	minimum real power output (MW)
PC1*	11	lower real power output of PQ capability curve (MW)
PC2*	12	upper real power output of PQ capability curve (MW)
$\mathtt{QC1MIN}^*$	13	minimum reactive power output at PC1 (MVAr)
QC1MAX*	14	maximum reactive power output at PC1 (MVAr)
QC2MIN*	15	minimum reactive power output at PC2 (MVAr)
QC2MAX*	16	maximum reactive power output at PC2 (MVAr)
$\mathtt{RAMP_AGC}^*$	17	ramp rate for load following/AGC (MW/min)
$\mathtt{RAMP_10}^*$	18	ramp rate for 10 minute reserves (MW)
RAMP_30*	19	ramp rate for 30 minute reserves (MW)
\mathtt{RAMP}_{Q}^*	20	ramp rate for reactive power (2 sec timescale) (MVAr/min)
\mathtt{APF}^*	21	area participation factor
MU_PMAX^{\dagger}	22	Kuhn-Tucker multiplier on upper P_q limit (u/MW)
MU_PMIN [†]	23	Kuhn-Tucker multiplier on lower P_q limit (u/MW)
MU_QMAX^\dagger	24	Kuhn-Tucker multiplier on upper Q_g limit $(u/MVAr)$
$\mathtt{MU}_{-}\mathtt{QMIN}^{\dagger}$	25	Kuhn-Tucker multiplier on lower Q_g limit (u/MVAr)

 $^{^*}$ Not included in version 1 case format.

 $^{^{\}dagger}$ Included in OPF output, typically not included (or ignored) in input matrix. Here we assume the

objective function has units u.

Used to determine voltage setpoint for optimal power flow only if opf.use_vg option is non-zero (0 by default). Otherwise generator voltage range is determined by limits set for corresponding bus in bus matrix.

Table B-3: Branch Data (mpc.branch)

name	column	description		
F_BUS	1	"from" bus number		
T_BUS	2	"to" bus number		
BR_R	3	resistance (p.u.)		
BR_X	4	reactance (p.u.)		
BR_B	5	total line charging susceptance (p.u.)		
$\mathtt{RATE_A}^*$	6	MVA rating A (long term rating), set to 0 for unlimited		
$\mathtt{RATE_B}^*$	7	MVA rating B (short term rating), set to 0 for unlimited		
$\mathtt{RATE_C}^*$	8	MVA rating C (emergency rating), set to 0 for unlimited		
TAP	9	transformer off nominal turns ratio, if non-zero (taps at "from"		
		bus, impedance at "to" bus, i.e. if $r = x = b = 0$, $tap = \frac{ V_f }{ V_f }$;		
		tap = 0 used to indicate transmission line rather than transformer,		
		i.e. mathematically equivalent to transformer with $tap = 1$)		
SHIFT	10	transformer phase shift angle (degrees), positive \Rightarrow delay		
BR_STATUS	11	initial branch status, $1 = \text{in-service}$, $0 = \text{out-of-service}$		
\mathtt{ANGMIN}^\dagger	12	minimum angle difference, $\theta_f - \theta_t$ (degrees)		
ANGMAX [†]	13	maximum angle difference, $\theta_f - \theta_t$ (degrees)		
PF [‡]	14	real power injected at "from" bus end (MW)		
QF [‡]	15	reactive power injected at "from" bus end (MVAr)		
PT [‡]	16	real power injected at "to" bus end (MW)		
QT [‡]	17	reactive power injected at "to" bus end (MVAr)		
MU_SF§	18	Kuhn-Tucker multiplier on MVA limit at "from" bus (u/MVA)		
MU_ST§	19	Kuhn-Tucker multiplier on MVA limit at "to" bus (u/MVA)		
MU_ANGMIN [§]	20	Kuhn-Tucker multiplier lower angle difference limit (u/degree)		
MU_ANGMAX [§]	21	Kuhn-Tucker multiplier upper angle difference limit (u/degree)		

^{*} Used to specify branch flow limits. By default these are limits on apparent power with units in MVA. However, the 'opf.flow_lim' option can be used to specify that the limits are active power or current, in which case the ratings are specified in MW or $(kA \cdot V_{basekV})$, respectively. For current this is equivalent to an MVA value at a 1 p.u. voltage.

[†] Not included in version 1 case format. The voltage angle difference is taken to be unbounded below if $\texttt{ANGMIN} \leq -360$ and unbounded above if $\texttt{ANGMAX} \geq 360$. If both parameters are zero, the voltage angle difference is unconstrained.

 $^{^{\}ddagger}$ Included in power flow and OPF output, ignored on input.

 $[\]S$ Included in OPF output, typically not included (or ignored) in input matrix. Here we assume the objective function has units u.

Table B-4: Generator Cost $Data^{\dagger}$ (mpc.gencost)

name	column	description
MODEL	1	cost model, 1 = piecewise linear, 2 = polynomial
STARTUP	2	startup cost in US dollars*
SHUTDOWN	3	shutdown cost in US dollars*
NCOST	4	number $N = n + 1$ of data points defining an n-segment piecewise linear cost
		function, or of coefficients defining an n -th order polynomial cost function
COST	5	parameters defining total cost function $f(p)$ begin in this column,
		units of f and p are f and MW (or MVAr), respectively
		$(\texttt{MODEL} = 1) \Rightarrow p_1, f_1, p_2, f_2, \dots, p_N, f_N$
		where $p_1 < p_2 < \cdots < p_N$ and the cost $f(p)$ is defined by
		the coordinates $(p_1, f_1), (p_2, f_2), \ldots, (p_N, f_N)$
		of the end/break-points of the piecewise linear cost
		$(\texttt{MODEL} = 2) \Rightarrow c_n, \dots, c_1, c_0$
		N coefficients of n -th order polynomial cost function, starting
		with highest order, where cost is $f(p) = c_n p^n + \cdots + c_1 p + c_0$

[†] If gen has n_g rows, then the first n_g rows of gencost contain the costs for active power produced by the corresponding generators. If gencost has $2n_g$ rows, then rows $n_g + 1$ through $2n_g$ contain the reactive power costs in the same format.

^{*} Not currently used by any Matpower functions.

Table B-5: DC Line Data* (mpc.dcline)

name	column	description
F_BUS	1	"from" bus number
T_BUS	2	"to" bus number
BR_STATUS	3	initial branch status, $1 = \text{in-service}$, $0 = \text{out-of-service}$
${ t PF}^{\dagger}$	4	real power flow at "from" bus end (MW), "from" \rightarrow "to"
\mathtt{PT}^\dagger	5	real power flow at "to" bus end (MW), "from" \rightarrow "to"
${\sf QF}^\dagger$	6	reactive power injected into "from" bus (MVAr)
$\mathtt{Q}\mathtt{T}^{\dagger}$	7	reactive power injected into "to" bus (MVAr)
VF	8	voltage magnitude setpoint at "from" bus (p.u.)
VT	9	voltage magnitude setpoint at "to" bus (p.u.)
PMIN	10	if positive (negative), lower limit on PF (PT)
PMAX	11	if positive (negative), upper limit on PF (PT)
QMINF	12	lower limit on reactive power injection into "from" bus (MVAr)
QMAXF	13	upper limit on reactive power injection into "from" bus (MVAr)
QMINT	14	lower limit on reactive power injection into "to" bus (MVAr)
QMAXT	15	upper limit on reactive power injection into "to" bus (MVAr)
LOSS0	16	coefficient l_0 of constant term of linear loss function (MW)
LOSS1	17	coefficient l_1 of linear term of linear loss function (MW/MW)
		$(p_{\text{loss}} = l_0 + l_1 p_f, \text{ where } p_f \text{ is the flow at the "from" end})$
MU_PMIN [‡]	18	Kuhn-Tucker multiplier on lower flow limit at "from" bus (u/MW)
$\mathtt{MU_PMAX}^\ddagger$	19	Kuhn-Tucker multiplier on upper flow limit at "from" bus (u/MW)
$\mathtt{MU}_{-}\mathtt{QMINF}^{\ddagger}$	20	Kuhn-Tucker multiplier on lower VAr limit at "from" bus $(u/MVAr)$
$\mathtt{MU}_{-}\mathtt{QMAXF}^{\ddagger}$	21	Kuhn-Tucker multiplier on upper VAr limit at "from" bus $(u/MVAr)$
$\mathtt{MU}_{-}\mathtt{QMINT}^{\ddagger}$	22	Kuhn-Tucker multiplier on lower VAr limit at "to" bus $(u/MVAr)$
MU_QMAXT [‡]	23	Kuhn-Tucker multiplier on upper VAr limit at "to" bus (u/MVAr)

^{*} Requires explicit use of toggle_dcline.

† Output column, value updated by power flow or OPF (except PF in case of simple power flow).

‡ Included in OPF output, typically not included (or ignored) in input matrix. Here we assume the objective function has units u.

Appendix C Matpower Options

Beginning with version 4.2, MATPOWER uses an options struct to control the many options available. Earlier versions used an options vector with named elements. MATPOWER's options are used to control things such as:

- power flow algorithm
- power flow termination criterion
- power flow options (e.g. enforcing of reactive power generation limits)
- continuation power flow options
- OPF algorithm
- OPF termination criterion
- OPF options (e.g. active vs. apparent power vs. current for line limits)
- verbose level
- printing of results
- solver specific options

As with the old-style options vector, the options struct should always be created and modified using the mpoption function to ensure compatibility across different versions of Matpower. The default Matpower options struct is obtained by calling mpoption with no arguments.

```
>> mpopt = mpoption;
```

Individual options can be overridden from their default values by calling mpoption with a set of name/value pairs as input arguments. For example, the following runs a fast-decoupled power flow of case30 with very verbose progress output:

```
>> mpopt = mpoption('pf.alg', 'FDXB', 'verbose', 3);
>> runpf('case30', mpopt);
```

For backward compatibility, old-style option names/values can also be used.

```
>> mpopt = mpoption('PF_ALG', 2, 'VERBOSE', 3);
```

Another way to specify option overrides is via a struct. Using the example above, the code would be as follows.

```
>> overrides = struct('pf', struct('alg', 'FDXB'), 'verbose', 3);
>> mpopt = mpoption(overrides);
```

Finally, a string containing the name of a function that returns such a struct, can be passed to mpoption instead of the struct itself.

```
>> mpopt = mpoption('verbose_fast_decoupled_pf_opts');
```

where the function verbose_fast_decoupled_pf_opts is defined as follows:

```
function ov = verbose_fast_decoupled_pf_opts()
ov = struct('pf', struct('alg', 'FDXB'), 'verbose', 3);
```

To make changes to an existing options struct (as opposed to the default options struct), simply include it as the first argument. For example, to modify the previous run to enforce reactive power limts, suppress the pretty-printing of the output and save the results to a struct instead:

```
>> mpopt = mpoption(mpopt, 'pf.enforce_q_lims', 1, 'out.all', 0);
>> results = runpf('case30', mpopt);
```

This works when specifying the overrides as a struct or function name as well. For backward compatibility, the first argument can be an old-style options vector, followed by old-style option name/value pairs.

The available options and their default values are summarized in the following tables and can also be accessed via the command help mpoption. Some of the options require separately installed optional packages available from the MATPOWER website.

Table C-1: Top-Level Options

name	default	description
model	'AC'	AC vs. DC modeling for power flow and OPF formulation
		'AC' – use AC formulation and corresponding algs/options 'DC' – use DC formulation and corresponding algs/options
pf	see Table C-2	power flow options
cpf	see Table C-3	continuation power flow options
opf	see Tables C-4, C-5	optimal power flow options
verbose	1	amount of progress info printed
	_	0 – print no progress info
		1 – print a little progress info
		2 – print a lot of progress info
		3 – print all progress info
out	see Table C-6	pretty-printed output options
mips	see Table C-7	MIPS options
clp	see Table C-8	CLP options*
cplex	see Table C-9	CPLEX options*
fmincon	see Table C-10	fmincon options †
glpk	see Table C-11	GLPK options*
gurobi	see Table C-12	Gurobi options*
ipopt	see Table C-13	IPOPT options*
knitro	see Table C-14	Artelys Knitro options*
minopf	see Table C-15	MINOPF options*
mosek	see Table C-16	MOSEK options*
pdipm	see Table C-17	PDIPM options*
tralm	see Table C-18	TRALM options*

^{*} Requires the installation of an optional package. See Appendix G for details on the corresponding package.
† Requires Matlab's Optimization Toolbox, available from The MathWorks, Inc (https://www.mathworks.com/).

Table C-2: Power Flow Options

name	default	description
pf.alg	'NR'	AC power flow algorithm: 'NR' - Newtons's method (formulation depends on values of pf.current_balance and pf.v_cartesian options) 'NR-SC' - Newtons's method (power mismatch, cartesian)† 'NR-IP' - Newtons's method (current mismatch, polar)† 'NR-IC' - Newtons's method (current mismatch, cartesian)† 'FDXB' - Fast-Decoupled (XB version) 'FDBX' - Fast-Decouple (BX version) 'GS' - Gauss-Seidel 'PQSUM' - Power Summation (radial networks only) 'ISUM' - Current Summation (radial networks only) 'YSUM' - Admittance Summation (radial networks only)
pf.current_balance [‡]	0	use current, as opposed to power, balance for AC PF, 0 or 1
pf.v_cartesian [‡]	0	use cartesian, as opposed to polar, representation for voltages for AC PF, 0 or 1
pf.tol	10^{-8}	termination tolerance on per unit P and Q dispatch
pf.nr.max_it	10	maximum number of iterations for Newton's method
pf.nr.lin_solver	20	linear solver option for mplinsolve for computing Newton update step (see mplinsolve for complete list of all options) '' - default to '\' for small systems, 'LU3' for larger ones '\' - built-in backslash operator 'LU' - explicit default LU decomposition and back substitution 'LU3' - 3 output arg form of lu, Gilbert-Peierls algorithm with approximate minimum degree (AMD) reordering 'LU4' - 4 output arg form of lu, UMFPACK solver (same as 'LU') 'LU5' - 5 output arg form of lu, UMFPACK solver w/row scaling
pf.fd.max_it	30	maximum number of iterations for fast-decoupled method
pf.gs.max_it	1000	maximum number of iterations for Gauss-Seidel method
pf.radial.max_it	20 0	maximum number of iterations for radial power flow methods
pf.radial.vcorr	U	perform voltage correction procedure in distribution power flow $0-\text{do }not$ perform voltage correction $1-\text{perform }voltage$ correction
pf.enforce_q_lims	0	enforce gen reactive power limits at expense of $ V_m $ 0 – do not enforce limits 1 – enforce limits, simultaneous bus type conversion 2 – enforce limits, one-at-a-time bus type conversion

[†] Automatically sets/overrides pf.current_balance and pf.v_cartesian options with corresponding values. ‡ Only relevant for Newton power flow solver.

Table C-3: Continuation Power Flow Options

name	default	description
cpf.parameterization	3	choice of parameterization
		1 — natural
		2 — arc length
		3 — pseudo arc length
${\tt cpf.stop_at}$	'NOSE'	determines stopping criterion
		'NOSE' — stop when nose point is reached
		'FULL' — trace full nose curve
		λ_{stop} — stop upon reaching target λ value λ_{stop}
${\tt cpf.enforce_p_lims}$	0	enforce gen active power limits
		0 — do <i>not</i> enforce limits
		1 — enforce limits
${\tt cpf.enforce_q_lims}$	0	enforce gen reactive power limits at expense of V_m
		0 — do not enforce limits
	0	1 — enforce limits
cpf.enforce_v_lims	0	enforce bus voltage magnitude limits
		0 — do not enforce limits
	0	1 — enforce limits
cpf.enforce_flow_lims	0	enforce branch MVA flow limits
		0 — do <i>not</i> enforce limits
	0.05	1 — enforce limits
cpf.step	0.05	default value for continuation power flow step size σ
cpf.step_min	10^{-4}	minimum allowed step size, σ_{\min}
cpf.step_max	0.2	maximum allowed step size, σ_{max}
cpf.adapt_step	0	toggle adaptive step size feature
		0 — adaptive step size disabled
enf adapt stop damping	0.7	1 — adaptive step size enabled
<pre>cpf.adapt_step_damping cpf.adapt_step_tol</pre>	10^{-3}	damping factor $\beta_{\rm cpf}$ from (5.13) for adaptive step sizing
cpf.target_lam_tol	10^{-5}	tolerance $\epsilon_{\rm cpf}$ from (5.13) for adaptive step sizing tolerance for target lambda detection
cpf.nose_tol	10^{-5}	tolerance for nose point detection (p.u.)
cpf.p_lims_tol	10^{-2}	tolerance for generator active power limit detection
ob: .b-::::::::::::::::::::::::::::::::::	10	(MW)
cpf.q_lims_tol	10^{-2}	tolerance for generator reactive power limit detection
-F4		(MVAr)
cpf.v_lims_tol	10^{-4}	tolerance for bus voltage magnitude limit detection (p.u)
cpf.flow_lims_tol	0.01	tolerance for branch flow limit detection (MVA)
cpf.plot.level	0	control plotting of nose curve
1 1	-	0 — do not plot nose curve
		1 — plot when completed
		2 — plot incrementally at each iteration
		3 — same as 2, with pause at each iteration
cpf.plot.bus	empty	index of bus whose voltage is to be plotted
cpf.user_callback	empty	string or cell array of strings with names of user callback
		functions [†]
		150

[†] See help cpf_default_callback for details.

Table C-4: OPF Solver Options

name	default	description
opf.ac.solver	'DEFAULT'	AC optimal power flow solver:
		'DEFAULT' - choose default solver, i.e. 'MIPS'
		'MIPS' - MIPS, MATPOWER Interior Point Solver,
		primal/dual interior point method [†]
		'FMINCON' - MATLAB Optimization Toolbox, fmincon
		'IPOPT' - IPOPT*
		'KNITRO' - Artelys Knitro*
		'MINOPF' - MINOPF, MINOS-based solver
		'PDIPM' - PDIPM, primal/dual interior point method [‡]
		'SDPOPF' - SDPOPF, solver based on semidefinite relax-
		ation
		'TRALM' - TRALM, trust region based augmented Lan-
		grangian method
opf.dc.solver	'DEFAULT'	DC optimal power flow solver:
		'DEFAULT' - choose default solver based on availability in the following order: 'GUROBI', 'CPLEX', 'MOSEK', 'OT', 'GLPK' (linear costs only),
		'BPMPD', 'MIPS'
		'MIPS' - MIPS, MATPOWER Interior Point Solver,
		primal/dual interior point method ^{\dagger}
		'BPMPD' - BPMPD*
		'CLP' - CLP*
		'CPLEX' - CPLEX*
		$\verb"GLPK" - GLPK" (no \ quadratic \ costs)$
		'GUROBI' - Gurobi*
		'IPOPT' - IPOPT*
		'MOSEK' - MOSEK*
		'OT' — MATLAB Opt Toolbox, quadprog, linprog

^{*} Requires the installation of an optional package. See Appendix G for details on the corresponding package.
† For MIPS-sc, the step-controlled version of this solver, the mips.step_control option must be set to 1.
‡ For SC-PDIPM, the step-controlled version of this solver, the pdipm.step_control option must be set to 1.

Table C-5: General OPF Options

name	default	description
opf.current_balance [‡]	0	use current, as opposed to power, balance formulation for AC OPF, 0 or 1
$\mathtt{opf.v_cartesian}^\ddagger$	0	use cartesian, as opposed to polar, representation for voltages for AC OPF, 0 or 1
opf.violation	5×10^{-6}	constraint violation tolerance
opf.use_vg	0	respect generator voltage setpoint, 0 or 1^{\dagger}
		 0 - use voltage magnitude limits specified in bus, ignore VG in gen 1 - replace voltage magnitude limits specified in bus by VG in corresponding gen
opf.flow_lim	'S'	quantity to limit for branch flow constraints
		'S' – apparent power flow (limit in MVA)
		'P' – active power flow (limit in MW)
		'I' – current magnitude (limit in MVA at 1 p.u.
		voltage, i.e $kA \cdot V_{\text{basekV}}$)
		'2' – same as 'P', but implemented using square
	0	of active flow, rather than simple max
opf.ignore_angle_lim	0	ignore angle difference limits for branches
		0 – include angle difference limits, if specified
onf goftling dofoult	1	1 – ignore angle difference limits even if specified default behavior of implicit soft limits, where parameters
opf.softlims.default	1	are not explicitly provided
		0 – do not include softlims if not explicitly specified
		1 – include softlims with default values if not ex-
		plicitly specified
opf.init_from_mpc*	-1	specify whether to use the current state in MATPOWER
open and and a		case to initialize OPF
		-1 – Matpower decides based on solver/algorithm
		0 – ignore current state in Matpower case [‡]
		1 – use current state in Matpower case
opf.start	0	strategy for initializing OPF starting point
		0 – default, Matpower decides based on solver,
		(currently identical to 1)
		1 – ignore current state in Matpower case [‡]
		2 – use current state in Matpower case
_	<u>^</u>	3 – solve power flow and use resulting state
opf.return_raw_der	0	for AC OPF, return constraint and derivative info in
		results.raw (in fields g, dg, df, d2f)

^{*} Deprecated. Use opf.start instead.
† Using a value between 0 and 1 results in the limits being determined by the corresponding weighted average of

[†] Only applies to fmincon, IPOPT, Artelys Knitro and MIPS solvers, which use an interior point estimate; others use current state in Matpower case, as with opf.start = 2.

Table C-6: Power Flow and OPF Output Options

name	default	description
verbose	1	amount of progress info to be printed
		0 – print no progress info
		1 – print a little progress info
		2 – print a lot of progress info
		3 – print all progress info
out.all	-1	controls pretty-printing of results
		-1 – individual flags control what is printed
		$0 - do \ not \ print \ anything^{\dagger*}$
		$1 - \text{print everything}^{\dagger}$
out.sys_sum	1	print system summary (0 or 1)
out.area_sum	0	print area summaries (0 or 1)
out.bus	1	print bus detail, includes per bus gen info (0 or 1)
out.branch	1	print branch detail (0 or 1)
out.gen	0	print generator detail (0 or 1)
out.lim.all	-1	controls constraint info output
		-1 – individual flags control what is printed
		$0-{ m do}\ not\ { m print\ any\ constraint\ info}^{\dagger}$
		$1 - \text{print only binding constraint info}^{\dagger}$
		$2 - \text{print all constraint info}^{\dagger}$
out.lim.v	1	control output of voltage limit info
		0 - do not print
		1 – print binding constraints only
		2 – print all constraints
out.lim.line	1	control output of line flow limit info [‡]
out.lim.pg	1	control output of gen active power limit info [‡]
$\mathtt{out.lim.qg}$	1	control output of gen reactive power limit info [‡]
out.force	0	print results even if success flag $= 0 (0 \text{ or } 1)$
${\tt out.suppress_detail}$	-1	suppress all output but system summary
		-1 – suppress details for large systems (> 500 buses)
		0 – do <i>not</i> suppress any output specified by other flags
		1 – suppress all output except system summary section [†]

^{*} This setting is ignored for pretty-printed output to files specified as FNAME argument in calls to runpf, runopf, etc.

† Overrides individual flags, but (in the case of out.suppress_detail) not out.all = 1.

† Takes values of 0, 1 or 2 as for out.lim.v.

Table C-7: OPF Options for MIPS

name	default	description
mips.linsolver	11	linear system solver for update step '' or '\' – built-in backslash \ operator
		'PARDISO' - PARDISO solver [†]
mips.feastol	0	feasibiliy (equality) tolerance
•		set to value of opf.violation by default
mips.gradtol	10^{-6}	gradient tolerance
mips.comptol	10^{-6}	complementarity condition (inequality) tolerance
mips.costtol	10^{-6}	optimality tolerance
mips.max_it	150	maximum number of iterations
mips.step_control	0	set to 1 to enable step-size control
$\texttt{mips.sc.red_it}^\ddagger$	20	maximum number of step size reductions per iteration

 $^{^\}dagger$ Requires installation of the optional PARDISO package. See Appendix G.11 for details.

Table C-8: OPF Options for ${\rm CLP}^{\dagger}$

name	default [§]	description	
clp.opts	empty	struct of native CLP options passed to clp_options to overrid	
${\tt clp.opt_fname}$	empty	defaults, applied after overrides from clp.opt_fname [‡] name of user-supplied function passed as FNAME argument to clp_options to override defaults [‡]	

 $^{^\}dagger$ For opf.dc.solver option set to 'CLP' only. Requires the installation of the optional CLP package. See Appendix G.2 for details.

† For details, see help clp.options or help clp.

 $^{^{\}ddagger}$ Only relevant when ${\tt mips.step_control}$ is on.

Table C-9: OPF Options for CPLEX †

name	default	description	
cplex.lpmethod	0	algorithm used by CPLEX for LP problems 0 – automatic; let CPLEX choose 1 – primal simplex 2 – dual simplex 3 – network simplex 4 – barrier 5 – sifting 6 – concurrent (dual, barrier, and primal)	
cplex.qpmethod	0	algorithm used by CPLEX for QP problems 0 – automatic; let CPLEX choose 1 – primal simplex 2 – dual simplex 3 – network simplex 4 – barrier	
cplex.opts	empty	struct of native CPLEX options (for cplexoptimset) passed to cplex_options to override defaults, applied after overrides from cplex.opt_fname [‡]	
cplex.opt_fname	empty	name of user-supplied function passed as FNAME argument to cplex_options to override defaults [‡]	
cplex.opt	0	if cplex.opt_fname is empty and cplex.opt is non-zero, the value of cplex.opt_fname is generated by appending cplex.opt to 'cplex_user_options_' (for backward compatibility with old MATPOWER option CPLEX_OPT)	

 $^{^\}dagger$ For opf.dc.solver option set to 'CPLEX' only. Requires the installation of the optional CPLEX package. See

Appendix G.3 for details.

‡ For details, see help cplex_options and the "Parameters of CPLEX" section of the CPLEX documentation at http://www.ibm.com/support/knowledgecenter/SSSA5P.

Table C-10: OPF Options for fmincon[†]

name	default	description	
fmincon.alg	4	algorithm used by fmincon in Matlab Opt Toolbox ≥ 4 1 - active-set [‡] 2 - interior-point, default "bfgs" Hessian approximation 3 - interior-point, "lbfgs" Hessian approximation 4 - interior-point, exact user-supplied Hessian 5 - interior-point, Hessian via finite-differences 6 - sop, sequential quadratic programming [‡]	
<pre>fmincon.tol_x fmincon.tol_f fmincon.max_it</pre>	$ \begin{array}{c} 10^{-4} \\ 10^{-4} \\ 0 \end{array} $	$6 - \text{sqp}$, sequential quadratic programming [‡] termination tolerance on x^* termination tolerance on f^* maximum number of iterations [*] $0 \Rightarrow \text{use solver's default value}$	

[†] Requires Matlab's Optimization Toolbox, available from The MathWorks, Inc (https://www.mathworks.com/).

Table C-11: OPF Options for $GLPK^{\dagger}$

name	$\mathrm{default}^\S$	description	
glpk.opts	empty	struct of native GLPK options passed to glpk_options to over	
glpk.opt_fname	empty	ride defaults, applied after overrides from glpk.opt_fname [‡] name of user-supplied function passed as FNAME argument to glpk_options to override defaults [‡]	

 $^{^{\}dagger}$ For opf.dc.solver option set to 'GLPK' only. Requires the installation of the optional GLPK package. See Appendix G.4 for details.

 $^{^{\}ddagger}$ Does not use sparse matrices, so not applicable for large-scale systems.

^{*} Display is set by verbose, TolCon by opf.violation, TolX by fmincon.tol_x, TolFun by fmincon.tol_f, and MaxIter and MaxFunEvals by fmincon.max_it.

[†] For details, see help glpk_options or the "param" section of the GLPK documentation at https://www.gnu.org/software/octave/doc/interpreter/Linear-Programming.html.

Table C-12: OPF Options for Gurobi[†]

name	$default^{\S}$	description			
gurobi.method	0	algorithm used by Gurobi for LP/QP problems			
		0 – primal simplex			
		1 – dual simplex			
		2 – barrier			
		3 – concurrent (LP only)			
		4 – deterministic concurrent (LP only)			
<pre>gurobi.timelimit</pre>	∞	maximum time allowed for solver (secs)			
gurobi.threads	0 (auto)	maximum number of threads to use			
gurobi.opts	empty	struct of native Gurobi options passed to gurobi_options to			
		override defaults, applied after overrides from gurobi.opt_fname [‡]			
${\tt gurobi.opt_fname}$	empty	name of user-supplied function passed as FNAME argument to			
		gurobi_options to override defaults [‡]			
gurobi.opt	0	if gurobi.opt_fname is empty and gurobi.opt is non-zer			
		the value of <pre>gurobi.opt_fname</pre> is generated by appending			
		<pre>gurobi.opt to 'gurobi_user_options_' (for backward compat-</pre>			
		ibility with old Matpower option GRB_OPT)			

 $^{^\}dagger$ For opf.dc.solver option set to 'GUROBI' only. Requires the installation of the optional Gurobi package. See Appendix G.5 for details.

Table C-13: OPF Options for ${\tt IPOPT}^\dagger$

name	default	description		
ipopt.opts	empty	struct of native IPOPT options (options.ipopt for ipopt) passed to ipopt_options to override defaults, applied after overrides from ipopt.opt_fname [‡]		
$ipopt.opt_fname$	empty	name of user-supplied function passed as FNAME argument to ipopt_options to override defaults [‡]		
ipopt.opt	0	if ipopt.opt_fname is empty and ipopt.opt is non-zero, the value of ipopt.opt_fname is generated by appending ipopt.opt to 'ipopt_user_options_' (for backward compatibility with old MATPOWER option IPOPT_OPT)		

[†] For opf.ac.solver or opf.dc.solver option set to 'IPOPT' only. Requires the installation of the optional IPOPT package [51]. See Appendix G.6 for details.

 $[\]S$ Default values in parenthesis refer to defaults assigned by Gurobi if called with option equal to 0.

For details, see help gurobi_options and the "Parameters" section of the "Gurobi Optimizer Reference Manual" at https://www.gurobi.com/documentation/.

For details, see help ipopt_options and the options reference section of the IPOPT documentation at http://www.coin-or.org/Ipopt/documentation/.

Table C-14: OPF Options for Artelys Knitro †

name	default	description	
knitro.tol_x	10^{-4}	termination tolerance on x	
knitro.tol_f	10^{-4}	termination tolerance on f	
knitro.maxit	0	maximum number of iterations	
		$0 \Rightarrow$ use solver's default value	
${\tt knitro.opt_fname}$	empty	name of user-supplied native Knitro options file that override	
		other options [‡]	
knitro.opt	0	if knitro.opt_fname is empty and knitro.opt is a posi-	
		tive integer n , the value of knitro.opt_fname is generated	
		as 'knitro_user_options_n.txt' (for backward compatibility	
		with old Matpower option KNITRO_OPT)	

[†] For opf.ac.solver option set to 'KNITRO' only. Requires the installation of the optional Artelys Knitro package [34]. See Appendix G.7 for details. ‡ Note that Artelys Knitro uses the opt_fname option slightly differently from other optional solvers. Specifically,

[‡] Note that Artelys Knitro uses the opt_fname option slightly differently from other optional solvers. Specifically, it is the name of a text file processed directly by Artelys Knitro, not a MATLAB function that returns an options struct passed to the solver.

Table C-15: OPF Options for MINOPF †

name	default [‡]	description		
minopf.feastol	$0 (10^{-3})$	primal feasibility tolerance		
		set to value of opf.violation by default		
minopf.rowtol	$0\ (10^{-3})$	row tolerance		
		set to value of opf.violation by default		
minopf.xtol	$0 (10^{-4})$	x tolerance		
minopf.majdamp	0(0.5)	major damping parameter		
minopf.mindamp	0(2.0)	minor damping parameter		
minopf.penalty	0(1.0)	penalty parameter		
${ t minopf.major_it}$	0(200)	major iterations		
${\tt minopf.minor_it}$	0(2500)	minor iterations		
${\tt minopf.max_it}$	0(2500)	iteration limit		
minopf.verbosity	-1	amount of progress output printed by MEX file		
		-1 – controlled by verbose option		
		0 - do not print anything		
		1 – print only only termination status message		
		2 – print termination status & screen progress		
		3 – print screen progress, report file (usually fort.9)		
minopf.core	0	memory allocation		
		defaults to $1200n_b + 2(n_b + n_g)^2$		
${\tt minopf.supbasic_lim}$	0	superbasics limit, defaults to $2n_b + 2n_g$		
minopf.mult_price	0 (30)	multiple price		

[†] For opf.ac.solver option set to 'MINOPF' only. Requires the installation of the optional MINOPF package [32]. See Appendix G.8 for details.

‡ Default values in parenthesis refer to defaults assigned in MEX file if called with option equal to 0.

Table C-16: OPF Options for $MOSEK^{\dagger}$

name	default [§]	description	
mosek.lp_alg	0	solution algorithm used by MOSEK for continuous LP problems (MSK_IPAR_OPTIMIZER)* 0 - automatic; let MOSEK choose 1 - interior point 3 - primal simplex 4 - dual simplex 5 - primal dual simplex 6 - automatic simplex (MOSEK chooses which simplex) 7 - network primal simplex 10 - concurrent	
${\tt mosek.max_it}$	0 (400)	<pre>interior point maximum iterations (MSK_IPAR_INTPNT_MAX_ITERATIONS)</pre>	
${\tt mosek.gap_tol}$	$0\ (10^{-8})$	interior point relative gap tolerance (MSK_DPAR_INTPNT_TOL_REL_GAP)	
${\tt mosek.max_time}$	0 (-1)	maximum time allowed for solver (negative means ∞) (MSK_DPAR_OPTIMIZER_MAX_TIME)	
${\tt mosek.num_threads}$	0 (1)	maximum number of threads to use (MSK_IPAR_INTPNT_NUM_THREADS)	
mosek.opts	empty	struct of native MOSEK options (param struct normally passed to mosekopt) passed to mosek_options to override defaults, applied after overrides from mosek.opt_fname [‡]	
${\tt mosek.opt_fname}$	empty	name of user-supplied function passed as FNAME argument to mosek_options to override defaults [‡]	
mosek.opt	0	if mosek.opt_fname is empty and mosek.opt is non-zero, the value of mosek.opt_fname is generated by appending mosek.opt to 'mosek_user_options_' (for backward compatibility with old MATPOWER option MOSEK_OPT)	

 $^{^\}dagger$ For opf.dc.solver option set to 'MOSEK' only. Requires the installation of the optional MOSEK package. See Appendix G.9 for details.

[§] Default values in parenthesis refer to defaults assigned by MOSEK if called with option equal to 0.

^{*} The values listed here correspond to those used by MOSEK version 7.x. Version 6.x was different. It is probably safer to write your code using the symbolic constants defined by mosek_symboon rather than using explicit numerical values.

[‡] For details, see help mosek_options and the "Parameters" reference in "The MOSEK optimization toolbox for MATLAB manual" at http://docs.mosek.com/7.1/toolbox/Parameters.html.

Table C-17: OPF Options for PDIPM †

name	default	description
pdipm.feastol	0	feasibiliy (equality) tolerance
		set to value of opf.violation by default
pdipm.gradtol	10^{-6}	gradient tolerance
pdipm.comptol	10^{-6}	complementarity condition (inequality) tolerance
pdipm.costtol	10^{-6}	optimality tolerance
${\tt pdipm.max_it}$	150	maximum number of iterations
$ t pdipm.step_control$	0	set to 1 to enable step-size control
${\tt pdipm.sc.red_it}^{\ddagger}$	20	maximum number of step size reductions per iteration
$ t pdipm.sc.smooth_ratio^{\ddagger}$	0.04	piecewise linear curve smoothing ratio

 $^{^\}dagger$ Requires installation of the optional TSPOPF package [28]. See Appendix G.13 for details. ‡ Only relevant when pdipm.step_control is on.

Table C-18: OPF Options for TRALM^\dagger

default	description
0	feasibiliy tolerance set to value of opf.violation by default
5×10^{-4}	primal variable tolerance
5×10^{-4}	dual variable tolerance
10^{-5}	optimality tolerance
40	maximum number of major iterations
100	maximum number of minor iterations
0.04	piecewise linear curve smoothing ratio
	$0 \\ 5 \times 10^{-4} \\ 5 \times 10^{-4} \\ 10^{-5} \\ 40 \\ 100$

 $^{^{\}dagger}$ Requires installation of the optional TSPOPF package [28]. See Appendix G.13 for details.

C.1 Mapping of Old-Style Options to New-Style Options

A MATPOWER options struct can be created from an old-style options vector simply by passing it to mpoption. The mapping of old-style options into the fields in the option struct are summarized in Table C-19. An old-style options vector can also be created from an options struct by calling mpoption with the struct and an empty second argument.

```
mpopt_struct = mpoption(mpopt_vector);
mpopt_vector = mpoption(mpopt_struct, []);
```

Table C-19: Old-Style to New-Style Option Mapping

idx	old option		new option	notes
2 3 4 5	PF_ALG		pf.alg 'NR' 'FDXB' 'FDBX' 'GS' pf.tol pf.nr.max_it pf.fd.max_it pf.gs.max_it	new option has string values
6 10	ENFORCE_Q_LIMS PF_DC 0 1	$\overset{\rightarrow}{\rightarrow}$	<pre>pf.enforce_q_lims model 'AC' 'DC'</pre>	new option has string values
11	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		opf.ac.solver 'DEFAULT' 'MINOPF' 'FMINCON' 'PDIPM' (and pdipm.step_con' 'PDIPM' (and pdipm.step_con' 'TRALM' 'MIPS' (and mips.step_contr' 'MIPS' (and mips.step_contr' 'IPOPT' 'KNITRO'	rol = 1)
16	OPF_VIOLATION		opf.violation	support for an at a bog been removed
17	CONSTR_TOL_X		<pre>fmincon.tol_x, knitro.tol_x</pre>	support for constr has been removed
18	CONSTR_TOL_F		<pre>fmincon.tol_f, knitro.tol_f</pre>	support for constr has been removed
19	CONSTR_MAX_IT		fmincon.max_it	support for constr has been removed

continued on next page

Table C-19: Old-Style to New-Style Option Mapping – continued

idx	old option	new option	notes
24	OPF_FLOW_LIM	opf.flow_lim	new option has string values
	$0 \rightarrow$	'S'	
	$1 \rightarrow$	'P'	
25	$ig 2 ightarrow 0$ PF_IGNORE_ANG_LIM	'I' opf.ignore_angle_lim	
$\frac{25}{26}$	OPF_ALG_DC	opf.dc.solver	new option has string values
20	0 →	'DEFAULT'	new option has string values
	$100 \rightarrow$		
	$200 \rightarrow$		
	$250 \rightarrow$	` 1	col = 1)
	$\begin{array}{ccc} 300 & \rightarrow \\ 400 & \rightarrow \end{array}$	'OT' 'IPOPT'	
	$500 \rightarrow$	'CPLEX'	
	$600 \rightarrow$	'MOSEK'	
	$ 700 \rightarrow$	'GUROBI'	
31	VERBOSE	verbose	
32	OUT_ALL	out.all	
33	OUT_SYS_SUM	out.sys_sum	
34	OUT_AREA_SUM	out.area_sum	
35	OUT_BUS	out.bug	
36	OUT_BRANCH	out.branch	
37	OUT_GEN	out.gen	
38	OUT_ALL_LIM	out.lim.all	
39	OUT_V_LIM	out.lim.v	
40	OUT_LINE_LIM	out.lim.line	
41	OUT_PG_LIM	out.lim.pg	
42	OUT_QG_LIM	out.lim.qg	
44	OUT_FORCE	out.force	
52	RETURN_RAW_DER	opf.return_raw_der	
55	FMC_ALG	fmincon.alg	
58	KNITRO_OPT	knitro.opt	
60	IPOPT_OPT	ipopt.opt	
61	MNS_FEASTOL	minopf.feastol	
62	MNS_ROWTOL	minopf.rowtol	
63	MNS_XTOL	minopf.xtol	
64	MNS_MAJDAMP	minopf.majdamp	
65	MNS_MINDAMP	minopf.mindamp	
66	MNS_PENALTY_PARM	minopf.penalty	
67	MNS_MAJOR_IT	minopf.major_it	

continued on next page

Table C-19: Old-Style to New-Style Option Mapping – continued

idx	old option	new option	notes
68	MNS_MINOR_IT	minopf.minor_it	
69	MNS_MAX_IT	${\tt minopf.max_it}$	
70	MNS_VERBOSITY	minopf.verbosity	
71	MNS_CORE	minopf.core	
72	MNS_SUPBASIC_LIM	minopf.supbasic_lim	
73	MNS_MULT_PRICE	${\tt minopf.mult_price}$	
80	FORCE_PC_EQ_PO	sopf.force_Pc_eq_P0	for c3sopf (not part of MATPOWER)
81	PDIPM_FEASTOL	mips.feastol,	
00	DD TD1/	pdipm.feastol	
82	PDIPM_GRADTOL	mips.gradtol,	
0.0	DD TDW GOWDEOI	pdipm.gradtol	
83	PDIPM_COMPTOL	mips.comptol,	
0.4	DDIDM GOGTTOI	pdipm.comptol	
84	PDIPM_COSTTOL	mips.costtol,	
95	DDTDM MAY TT	pdipm.costtol	
85	PDIPM_MAX_IT	mips.max_it,	
86	SCPDIPM_RED_IT	<pre>pdipm.max_it mips.sc.red_it,</pre>	
80	204014M-VED-11	pdipm.sc.red_it	
		paipm.sc.rea_rc	
87	TRALM_FEASTOL	tralm.feastol	
88	TRALM_PRIMETOL	tralm.primaltol	
89	TRALM_DUALTOL	tralm.dualtol	
90	TRALM_COSTTOL	tralm.costtol	
91	TRALM_MAJOR_IT	tralm.major_it	
92	TRALM_MINOR_IT	${ t tralm.minor_it}$	
93	SMOOTHING_RATIO	tralm.smooth_ratio,	
		pdipm.sc.smooth_ratio	
95	CPLEX_LPMETHOD	cplex.lpmethod	
98	CPLEX_QPMETHOD	cplex.qpmethod	
97	CPLEX_OPT	cplex.opt	
111	MOSEK_LP_ALG	$mosek.lp_alg$	
112	MOSEK_MAX_IT	${\tt mosek.max_it}$	
113	MOSEK_GAP_TOL	${\tt mosek.gap_tol}$	
114	MOSEK_MAX_TIME	${\tt mosek.max_time}$	
115	MOSEK_NUM_THREADS	${\tt mosek.num_threads}$	
116	MOSEK_OPT	mosek.opt	

continued on next page

Table C-19: Old-Style to New-Style Option Mapping – continued

idx	old option	new option	notes
121 122 123 124	GRB_METHOD GRB_TIMELIMIT GRB_THREADS GRB_OPT	<pre>gurobi.method gurobi.timelimit gurobi.threads gurobi.opt</pre>	

Appendix D Matpower Files and Functions

This appendix lists all of the files and functions that MATPOWER provides, with the exception of those in the extras directory (see Appendix E). In most cases, the function is found in a MATLAB M-file of the same name in the lib directory of the distribution, where the .m extension is omitted from this listing. For more information on each, at the MATLAB prompt, simply type help followed by the name of the function. For documentation and data files, the filename extensions are included.

D.1 Directory Layout and Documentation Files

Some of these files appear at the top level of the distribution and others in the docs directory.

Table D-1: MATPOWER Directory Layout and Documentation Files

name	description
data/	Matpower cases (see Section D.3)
docker/	files related to MATPOWER Docker image
Dockerfile	commands for building Docker image
MATPOWER-Docker.md	documentation for MATPOWER Docker image
docker-compose.test.yml	configuration for Docker Hub auto test
matpower_desktop.py	Python script for use with MATPOWER Docker
matpower_docker_tests.sh	script for Docker Hub auto test
docs/	Matpower documentation
MATPOWER-dev-guide.md	Matpower Developer's Guide
MATPOWER-manual.pdf	Matpower User's Manual [†]
relnotes/	Matpower release notes
src/	I⁴T _E X source for Matpower User's Manual
TN1-OPF-Auctions.pdf	MATPOWER Technical Note 1 "Uniform Price
	Auctions and Optimal Power Flow" [48]
TN2-OPF-Derivatives.pdf	MATPOWER Technical Note 2 "AC Power Flows,
	Generalized OPF Costs and their Derivatives us-
	ing Complex Matrix Notation" [38]
TN3-More-OPF-Derivatives.pdf	MATPOWER Technical Note 3 "Addendum to AC
	Power Flows and their Derivatives using Complex
	Matrix Notation: Nodal Current Balance" [39]
TN4-OPF-Derivatives-Cartesian.pdf	MATPOWER Technical Note 4 "AC Power Flows
	and their Derivatives using Complex Matrix No-
	tation and Cartesian Coordinate Voltages" [40]
lib/	Matpower software (see Section D.2)
caseformat.m	help file documenting Matpower case for-
	mat (i.e. at the command prompt, type help
	caseformat to see details)
t/	MATPOWER tests and examples (see Section D.4)
mips/	integrated MIPS distribution [†]
most/	integrated MOST distribution [†]
mptest/	integrated MP-Test distribution (see Table D-24)
AUTHORS	list of Matpower authors
CHANGES.md	MATPOWER change history
CITATION	info on how to cite MATPOWER
CONTRIBUTING.md	Matpower Contributors Guide
LICENSE	license file
README.md	basic introduction to Matpower
install_matpower.m	Matpower installer script

[†] The MIPS User's Manual and MOST User's Manual are found in <matpower>/mips/docs and <matpower>/most/docs, respectively.

Matpower Functions D.2

All of the functions listed in this section are found in <MATPOWER>/lib, unless noted otherwise.

Table D-2: Top-Level Simulation Functions

name	description
runpf	power flow [†]
runcpf	AC continuation power flow
runopf	optimal power flow [†]
runuopf	optimal power flow with unit-decommitment [†]
rundcpf	DC power flow [‡]
rundcopf	DC optimal power flow [‡]
runduopf	DC optimal power flow with unit-decommitment [‡]
runopf_w_res	optimal power flow with fixed reserve requirements [†]
most	MOST, MATPOWER Optimal Scheduling Tool §

Table D-3: Input/Output Functions

name	description
cdf2mpc	converts power flow data from IEEE Common Data Format (CDF) to MATPOWER format
loadcase	loads data from a MATPOWER case file or struct into data matrices or a case struct
mpoption	sets and retrieves Matpower options
printpf	pretty prints power flow and OPF results
psse2mpc	converts power flow data from PSS/E RAW format to MATPOWER format
save2psse	exports Matpower case to PSS/E RAW format
savecase	saves case data to a Matpower case file

 $^{^\}dagger$ Uses AC model by default. ‡ Simple wrapper function to set option to use DC model before calling the corresponding general function above.

 $[\]S$ MOST and its supporting files and functions in the most/ sub-directory are documented in the MOST User's Manual and listed in its Appendix A.

Table D-4: Data Conversion Functions

name	description
ext2int e2i_data e2i_field int2ext i2e_data i2e_field get_reorder set_reorder	converts case from external to internal indexing converts arbitrary data from external to internal indexing converts fields in mpc from external to internal indexing converts case from internal to external indexing converts arbitrary data from internal to external indexing converts fields in mpc from internal to external indexing returns A with one of its dimensions indexed assigns B to A with one of the dimensions of A indexed

Table D-5: Power Flow Functions

name	description
calc_v_i_sum	implementation of current summation power flow solver [†]
$calc_v_pq_sum$	implementation of power summation power flow solver [†]
$calc_v_y_sum$	implementation of admittance summation power flow solver [†]
dcpf	implementation of DC power flow solver
fdpf	implementation of fast-decoupled power flow solver
gausspf	implementation of Gauss-Seidel power flow solver
${\tt make_vcorr}$	calculate voltage corrections for radial power flow [†]
$make_zpv$	form loop impedance for PV bus handling in radial power flow [†]
newtonpf	implementation of Newton-method power flow solver (power, polar)
${\tt newtonpf_I_cart}$	implementation of Newton-method power flow solver (current, cartesian)
$newtonpf_I_polar$	implementation of Newton-method power flow solver (current, polar)
$newtonpf_S_cart$	implementation of Newton-method power flow solver (power, cartesian)
$order_radial$	performs oriented ordering of buses and branches for radial power flow [†]
pfsoln	computes branch flows, generator reactive power (and real power for
	slack bus), updates bus, gen, branch matrices with solved values
radial_pf	wrapper for backward/forward sweep power flow solvers [†]

 $^{^{\}dagger}$ For radial networks only.

Table D-6: Continuation Power Flow Functions

name	description
cpf_corrector	computes Newton method corrector steps
cpf_current_mpc	construct Matpower case struct for current continuation step
cpf_default_callback	callback function that accumulates, and optionally plots, results from each iteration
cpf_detect_events	detects event intervals and zeros given previous and current event function values
cpf_flim_event_cb	callback function to handle 'FLIM' events
cpf_flim_event	event function for 'FLIM' events, to detect branch MVA flow limits
cpf_nose_event_cb	callback function to handle 'NOSE' events
${\tt cpf_nose_event}$	event function for 'NOSE' events, to detect nose point of continuation curve
$\mathtt{cpf}_{-}\mathtt{p}_{-}\mathtt{jac}$	computes partial derivatives of parameterization function
$\mathtt{cpf}_{-\!p}$	computes value of parameterization function at current solution
cpf_plim_event_cb	callback function to handle 'PLIM' events
cpf_plim_event	event function for 'PLIM' events, to detect generator active power limits
cpf_predictor	computes the predicted solution from the current solution and tangent direction
${\tt cpf_qlim_event_cb}$	callback function to handle 'QLIM' events
cpf_qlim_event	event function for 'QLIM' events, to detect generator reactive power limits
cpf_register_callback	registers a CPF callback function to be called by runcpf
cpf_register_event	registers a CPF event function to be called by runcpf
$\mathtt{cpf}_{\mathtt{_tangent}}$	computes the normalized tangent vector at the current solution
cpf_target_lam_event_cb	callback function to handle 'TARGET_LAM' events
cpf_target_lam_event	event function for 'TARGET_LAM' events, to detect end of continuation curve or other target λ value
cpf_vlim_event_cb	callback function to handle 'VLIM' events
cpf_vlim_event	event function for 'VLIM' events, to detect bus voltage magnitude limits

Table D-7: OPF and Wrapper Functions

name	description
opf [†] dcopf [‡] fmincopf [‡] mopf [‡] uopf [‡]	the main OPF function, called by runopf calls opf with options set to solve DC OPF calls opf with options set to use fmincon to solve AC OPF calls opf with options set to use MINOPF to solve AC OPF implements unit-decommitment heuristic, called by runuopf

[†] Can also be used as a top-level function, run directly from the command line. It provides more calling options than runopf, primarly for backward compatibility with previous versions of mopf from MINOPF, but does not offer the option to save the output or the solved case.

[‡] Wrapper with same calling conventions as opf.

 $[\]S$ Requires the installation of an optional package. See Appendix ${\color{red}G}$ for details on the corresponding package.

Table D-8: OPF Model Objects

name	description
@opf_model/	OPF model object used to encapsulate OPF problem formulation
display	called to display object when statement not ended with semicolon
get_mpc	returns the Matpower case struct
opf_model	constructor for the opf_model class
<pre>@opt_model/</pre>	optimization model object (@opf_model base class)
${\tt add_legacy_cost}$	adds a named subset of legacy user-defined costs to the model
$\mathtt{add_lin_constraint}$	adds a named subset of linear constraints to the model
$\verb"add_named_set"^\dagger$	adds a named subset of costs, constraints or variables to the model
${\tt add_nln_constraint}$	adds a named subset of nonlinear constraints to the model
add_nln_cost	adds a named subset of general nonlinear costs to the model
add_quad_cost	adds a named subset of quadratic costs to the model
add_var	adds a named subset of optimization variables to the model
$describe_idx$	identifies variable, constraint or cost row indices
display	called to display object when statement not ended with semicolon
${\tt eval_legacy_cost}$	evaluates legacy user costs and derivatives
$\verb eval_nln_constraint \\$	returns full set of nonlinear equality or inequality constraints and
	their gradients
eval_nln_constraint_hess	returns Hessian for full set of nonlinear equality or inequality con-
	straints
eval_nln_cost	evaluates general nonlinear costs and derivatives
$\verb eval_quad_cost $	evaluates quadratic costs and derivatives
$\mathtt{get_idx}$	returns the idx struct for vars, lin/nln constraints, costs
$\mathtt{get}_\mathtt{userdata}$	returns values of user data stored in the model
${ t getN}$	returns the number of variables, constraints or cost rows [‡]
get	returns the value of a field of the object
${ t init_indexed_name}$	initializes dimensions for indexed named set of costs, constraints
	or variables
opt_model	constructor for the opt_model class
${\tt params_lin_constraint}$	returns individual or full set of linear constraint parameters
$params_legacy_cost$	returns individual or full set of legacy user cost coefficients
$params_nln_cost$	returns individual general nonlinear cost parameters
$params_quad_cost$	returns individual or full set of quadratic cost coefficients
params_var	returns initial values, bounds and variable type for optimimization vector \hat{x}^{\ddagger}
${\tt valid_named_set_type}^{\dagger}$	returns label for given named set type if valid, empty otherwise
${\tt varsets_cell2struct}^{\dagger}$	converts variable set list from cell array to struct array
$varsets_idx$	returns vector of indices into opt vector \hat{x} for variable set list
varsets_len	returns total number of optimization variables for variable set list
$varsets_x$	assembles cell array of sub-vectors of opt vector \hat{x} specified by variable set list

[†] Private method for internal use only. ‡ For all, or alternatively, only for a named (and possibly indexed) subset.

Table D-9: Deprecated Copt_model Methods

name	description
@opt_model/ add_constraints add_costs add_vars build_cost_params compute_cost get_cost_params getv linear_constraints	optimization model object (@opf_model base class) use add_lin_constraint or add_nln_constraint instead† use add_legacy_cost, add_nln_cost or add_quad_cost instead† use add_var instead† no longer needed use eval_legacy_cost instead use params_legacy_cost instead use params_var instead use params_lin_constraint instead

 $^{^{\}dagger}$ Or ${\tt init_indexed_name}$ if simply initializing the dimensions for an indexed named set.

Table D-10: OPF Solver Functions

name	description
dcopf_solver	sets up and solves DC OPF problem
$fmincopf_solver$	sets up and solves OPF problem using fmincon, Matlab Opt Toolbox
$ipoptopf_solver$	sets up and solves OPF problem using IPOPT [†]
ktropf_solver	sets up and solves OPF problem using Artelys Knitro [†]
$mipsopf_solver$	sets up and solves OPF problem using MIPS
mopf_solver	sets up and solves OPF problem using MINOPF [†]
tspopf_solver	sets up and solves OPF problem using PDIPM, SC-PDIPM or TRALM [†]

 $^{^{\}dagger}$ Requires the installation of an optional package. See Appendix $^{\mathbf{G}}$ for details on the corresponding package.

Table D-11: Other OPF Functions

name	description
margcost	computes the marginal cost of generation as a function of generator output
${\tt makeAang}$	forms linear constraints for branch angle difference limits
${\tt makeApq}$	forms linear constraints for generator PQ capability curves
makeAvl	forms linear constraints for dispatchable load constant power factor
${\tt makeAy}$	forms linear constraints for piecewise linear generator costs (CCV)
opf_args	input argument handling for opf
${\tt opf_setup}$	constructs an OPF model object from a MATPOWER case
opf_execute	executes the OPF specified by an OPF model object
opf_branch_ang_fcn [†]	evaluates AC branch angle difference limit constraints and gradients [‡]
opf_branch_ang_hess [†]	evaluates Hessian of AC branch angle difference limit constraints [‡]
opf_branch_flow_fcn [†]	evaluates AC branch flow limit constraints and gradients
$ exttt{opf_branch_flow_hess}^\dagger$	evaluates Hessian of AC branch flow limit constraints
${\tt opf_consfcn}^{\dagger}$	evaluates function and gradients for AC OPF nonlinear con-
	straints
${\tt opf_costfcn}^\dagger$	evaluates function, gradients and Hessian for AC OPF objective function
${\tt opf_current_balance_fcn}^\dagger$	evaluates AC current balance constraints and gradients
opf_current_balance_hess [†]	evaluates Hessian of AC current balance constraints
${\tt opf_gen_cost_fcn}^\dagger$	evaluates polynomial generator costs and derivatives
opf_hessfcn [†]	evaluates the Hessian of the Lagrangian for AC OPF
opf_legacy_user_cost_fcn [†]	evaluates legacy user-defined costs and derivatives
opf_power_balance_fcn [†]	evaluates AC power balance constraints and gradients
opf_power_balance_hess [†]	evaluates Hessian of AC power balance constraints
opf_veq_fcn [†]	evaluates voltage magnitude equality constraints and gradients [‡]
${\sf opf_veq_hess}^\dagger$	evaluates Hessian of voltage magnitude equality constraints [‡]
opf_vlim_fcn [†]	evaluates voltage magnitude limit constraints and gradients [‡]
opf_vlim_hess [†]	evaluates Hessian of voltage magnitude limit constraints [‡]
opf_vref_fcn [†]	evaluates reference voltage angle equality constraints and $\operatorname{gradients}^{\ddagger}$
opf_vref_hess [†]	evaluates Hessian of reference voltage angle equality constraints [‡]
totcost	computes the total cost of generation as a function of generator
	output
update_mupq	updates generator limit prices based on the shadow prices on capability curve constraints

[†] Used by fmincon, MIPS, IPOPT and Artelys Knitro for AC OPF.

† Used with cartesian coordinate voltage representation only. I.e. opf.v_cartesian option set to 1.

Table D-12: OPF User Callback Functions

name	description
add_userfcn	appends a user callback function to the list of those to be called for a given case
$remove_userfcn$	removes a user callback function from the list
$run_userfcn$	executes the user callback functions for a given stage
${ t toggle_dcline}$	enable/disable or check the status of the callbacks implementing DC lines
$toggle_{_}iflims$	enable/disable or check the status of the callbacks implementing interface flow limits
toggle_reserves	enable/disable or check the status of the callbacks implementing fixed reserve requirements
$toggle_softlims$	enable/disable or check the status of the callbacks implementing DC OPF branch flow soft limits

Table D-13: Power Flow Derivative Functions

name	${ m description}^\dagger$
dIbr_dV	evaluates the partial derivatives of $I_{f t}$ with respect to V
$dSbr_{-}dV$	evaluates the partial derivatives of $S_{f t}$ with respect to V
dAbr_dV	evaluates the partial derivatives of $ F_{f t} ^2$ with respect to V
${\tt dImis_dV}$	evaluates the partial derivatives of I_{mis} with respect to V
$dSbus_dV$	evaluates the partial derivatives of $S_{\rm bus}$ with respect to V
$d2Ibr_dV2$	evaluates the 2^{nd} derivatives of $I_{f t}$ with respect to V
$d2Sbr_dV2$	evaluates the 2 nd derivatives of $S_{f t}$ with respect to V
$d2Abr_dV2$	evaluates the 2 nd derivatives of $ \tilde{F}_{f t} ^2$ with respect to V
$\mathtt{d2AIbr_dV2}^*$	evaluates the 2 nd derivatives of $ I_{f t} ^2$ with respect to V
$\mathtt{d2ASbr_dV2}^*$	evaluates the 2 nd derivatives of $ S_{f t} ^2$ with respect to V
$d2Imis_dV2$	evaluates the $2^{\rm nd}$ derivatives of $I_{\rm mis}$ with respect to V
$d2Imis_dVdSg$	evaluates the $2^{\rm nd}$ derivatives of $I_{\rm mis}$ with respect to V and S_q .
d2Sbus_dV2	evaluates the 2^{nd} derivatives of S_{bus} with respect to V

Deprecated. Please use d2Abr_dV2 instead.

Deprecated. Please use d2Abr.4v2 instead.

† V represents complex bus voltages, S_g the complex generator power injections, $I_{f|t}$ complex branch current injections, $S_{f|t}$ complex branch power injections, I_{mis} complex bus current injections, S_{bus} complex bus power injections and $F_{f|t}$ refers to branch flows, either $I_{f|t}$ or $S_{f|t}$, depending on the inputs. The second derivatives are all actually partial derivatives of the product of a first derivative matrix and a vector λ . Please see Matrower Technical Note 2 [38], Matrower Technical Note 3 [39] and Matrower Technical Note 4 [40] for the latest of the formulas implemented by those functions the detailed derivations of the formulas implemented by these functions.

Table D-14: NLP, LP & QP Solver Functions

name	description
${\tt clp_options}$	default options for CLP solver [†]
${\tt cplex_options}$	default options for CPLEX solver [†]
${ t glpk_options}$	default options for GLPK solver [†]
gurobi_options	default options for Gurobi solver [†]
gurobiver	prints version information for Gurobi/Gurobi_MEX
$ipopt_options$	default options for IPOPT solver [†]
mips	MATPOWER Interior Point Solver – primal/dual interior point solver for NLP
mipsver	prints version information for MIPS
${\tt miqps_matpower}$	Mixed-Integer Quadratic Program Solver for Matpower, wrapper function
	provides a common MIQP solver interface for various MIQP/MILP solvers
${\tt miqps_cplex}$	common MIQP/MILP solver interface to CPLEX (cplexmiqp and
	${\tt cplexmilp})^{\dagger}$
${\tt miqps_glpk}$	common MILP solver interface to GLPK [†]
${\tt miqps_gurobi}$	common MIQP/MILP solver interface to Gurobi [†]
${\tt miqps_mosek}$	common MIQP/MILP solver interface to MOSEK (mosekopt) [†]
${\tt miqps_ot}$	common QP/MILP solver interface to MATLAB Opt Toolbox's intlinprog,
	quadprog, linprog
${\tt mosek_options}$	default options for MOSEK solver [†]
${\tt mosek_symbcon}$	symbolic constants to use for MOSEK solver options [†]
mplinsolve	common linear system solver interface, used by MIPS
mpopt2qpopt	create mi/qps_matpower options struct from MATPOWER options struct
${\tt qps_matpower}$	Quadratic Program Solver for Matpower, wrapper function
	provides a common QP solver interface for various QP/LP solvers
$\mathtt{qps_bpmpd}$	common QP/LP solver interface to BPMPD_MEX [†]
${\tt qps_clp}$	common QP/LP solver interface to CLP †
${\tt qps_cplex}$	common QP/LP solver interface to CPLEX (cplexqp and cplexlp) †
$\mathtt{qps_glpk}$	common QP/LP solver interface to GLPK [†]
qps_gurobi	common QP/LP solver interface to Gurobi [†]
${ t qps_ipopt}$	common QP/LP solver interface to IPOPT-based solver [†]
qps_mips	common QP/LP solver interface to MIPS-based solver
qps_mosek	common QP/LP solver interface to MOSEK $(mosekopt)^{\dagger}$
$\mathtt{qps_ot}$	common QP/LP solver interface to MATLAB Opt Toolbox's quadprog,
	linprog

 $^{^{\}dagger}$ Requires the installation of an optional package. See Appendix ${}^{\mathbf{G}}$ for details on the corresponding package.

Table D-15: Matrix Building Functions

name	description
makeB	forms the fast-decoupled power flow matrices, B^\prime and $B^{\prime\prime}$
makeBdc	forms the system matrices B_{bus} and B_f and vectors $P_{f,\text{shift}}$ and $P_{\text{bus,shift}}$
	for the DC power flow model
${ t makeJac}$	forms the power flow Jacobian matrix
makeLODF	forms the line outage distribution factor matrix
makePTDF	forms the DC PTDF matrix for a given choice of slack
makeSbus	forms the vector of complex bus power injections
${\tt makeSdzip}$	forms a struct w/vectors of nominal values for each component of ZIP load
makeYbus	forms the complex bus and branch admittance matrices Y_{bus},Y_f and Y_t

Table D-16: Utility Functions

name	description
apply_changes	modifies an existing Matpower case by applying changes defined
	in a change table
bustypes	creates vectors of bus indices for reference bus, PV buses, PQ buses
calc_branch_angle	calcultes a vector of voltage angle differences for branches
case_info	checks a Matpower case for connectivity and prints a system
compare_case	summary prints summary of differences between two MATPOWER cases
define_constants	convenience script defines constants for named column indices to
dollino_comp can op	data matrices (calls idx_bus, idx_brch, idx_gen and idx_cost)
extract_islands	extracts islands in a network into their own MATPOWER case struct
feval_w_path	same as MATLAB's feval function, except the function being eval-
	uated need not be in the MATLAB path
find_islands	finds islands and isolated buses in a network
genfuels	list of standard values for generator fuel types, for mpc.genfuel
gentypes	list of standard values for generator unit types, for mpc.gentype
get_losses	compute branch losses (and derivatives) as functions of bus voltage
hasPQcap	checks for generator P-Q capability curve constraints
have_fcn	checks for availability of optional functionality
idx_brch	named column index definitions for branch matrix
idx_bus	named column index definitions for bus matrix
idx_cost	named column index definitions for gencost matrix
idx_ct	constant definitions for use with <i>change tables</i> and apply_changes
idx_dcline	named column index definitions for dcline matrix
idx_gen	named column index definitions for gen matrix
isload	checks if generators are actually dispatchable loads
load2disp	converts fixed loads to dispatchable loads
loadshed	computes MW curtailments of dispatchable loads
mpver	prints version information for Matpower and optional packages
modcost	modifies gencost by horizontal or vertical scaling or shifting
nested_struct_copy	copies the contents of nested structs
poly2pwl	creates piecewise linear approximation to polynomial cost function
polycost	evaluates polynomial generator cost and its derivatives
pqcost	splits gencost into real and reactive power costs
savechgtab	saves a change table, such as used by apply_changes, to a file
scale_load	scales fixed and/or dispatchable loads by load zone
total_load	returns vector of total load in each load zone

Table D-17: Other Functions

name	description
connected_components	returns the connected components of a graph
${\tt mpoption_info_clp}$	option information for CLP
mpoption_info_cplex	option information for CPLEX
mpoption_info_fmincon	option information for FMINCON
mpoption_info_glpk	option information for GLPK
mpoption_info_gurobi	option information for Gurobi
mpoption_info_intlinprog	option information for INTLINPROG
mpoption_info_ipopt	option information for IPOPT
mpoption_info_knitro	option information for Artelys Knitro
mpoption_info_linprog	option information for LINPROG
mpoption_info_mips	option information for MIPS
${\tt mpoption_info_mosek}$	option information for MOSEK
mpoption_info_quadprog	option information for QUADPROG
psse_convert	converts data read from PSS/E RAW file to MATPOWER format
psse_convert_hvdc	called by psse_convert to handle HVDC data
psse_convert_xfmr	called by psse_convert to handle transformer data
psse_parse	parses data from a PSS/E RAW data file
psse_parse_line	called by psse_parse to parse a single line
${\tt psse_parse_section}$	called by psse_parse to parse an entire section
psse_read	reads data from a PSS/E RAW data file

D.3 Example Matpower Cases

All of the functions listed in this section are found in <MATPOWER>/data, unless noted otherwise.

Table D-18: Small Transmission System Test Cases

name	description
case4gs	4-bus example case from Grainger & Stevenson
case5	modified 5-bus PJM example case from Rui Bo
case6ww	6-bus example case from Wood & Wollenberg
case9	9-bus example case from Chow
case9Q	case9 with reactive power costs
case9target	modified case9, target for example continuation power flow
case14	IEEE 14-bus case
case24_ieee_rts	IEEE RTS 24-bus case
case30	30-bus case, based on IEEE 30-bus case
case_ieee30	IEEE 30-bus case
case30pwl	case30 with piecewise linear costs
case30Q	case30 with reactive power costs
case39	39-bus New England case
case57	IEEE 57-bus case
case_RTS_GMLC	96-machine, 73-bus Reliability Test System*
case118	IEEE 118-bus case
case145	IEEE 145-bus case, 50 generator dynamic test case
case300	IEEE 300-bus case

^{*} Reliability Test System Grid Modernization Lab Consortium data from https://github.com/GridMod/RTS-GMLC.

Table D-19: Small Radial Distribution System Test Cases

name	description
case4_dist case18 case22 case33bw case69 case85 case141	4-bus example radial distribution system case 18-bus radial distribution system from Grady, Samotyj and Noyola 22-bus radial distribution system from Raju, Murthy and Ravindra 33-bus radial distribution system from Baran and Wu 69-bus radial distribution system from Das 85-bus radial distribution system from Das, Kothari and Kalam 141-bus radial distribution system from Khodr, Olsina, De Jesus and Yusta

Table D-20: ACTIV Synthetic Grid Test Cases

name	description
case_ACTIVSg200	200-bus Illinois synthetic model*
case_ACTIVSg500	500-bus South Carolina synthetic model*
case_ACTIVSg2000	2000-bus Texas synthetic model*
case_ACTIVSg10k	10,000-bus US WECC synthetic model*
case_ACTIVSg25k	25,000-bus US Northeast/Mid-Atlantic synthetic model*
case_ACTIVSg70k	70,000-bus Eastern US synthetic model*
${\tt case_SyntheticUSA}$	82,000-bus continental USA synthetic model*
contab_ACTIVSg200	contingency table for case_ACTIVSg200*
contab_ACTIVSg500	contingency table for case_ACTIVSg500*
contab_ACTIVSg2000	contingency table for case_ACTIVSg2000*
contab_ACTIVSg10k	contingency table for case_ACTIVSg10k*
scenarios_ACTIVSg200	one year of hourly zonal load scenarios for case_ACTIVSg200*
scenarios_ACTIVSg2000	one year of hourly area load scenarios for ${\tt case_ACTIVSg2000}^*$

These synthetic grid models are fictitious representations that are designed to be statistically and functionally similar to actual electric grids while containing no confidential critical energy infrastructure information (CEII). Please cite reference [44] when publishing results based on this data. More information about these synthetic cases can be found at https://electricgrids.engr.tamu.edu.

Table D-21: Polish System Test Cases

name	description
case2383wp case2736sp case2737sop case2746wop case2746wp	Polish system - winter 1999-2000 peak* Polish system - summer 2004 peak* Polish system - summer 2004 off-peak* Polish system - winter 2003-04 off-peak* Polish system - winter 2003-04 evening peak*
case3012wp case3120sp case3375wp	Polish system - winter 2007-08 evening peak* Polish system - summer 2008 morning peak* Polish system plus - winter 2007-08 evening peak*

^{*} Contributed by Roman Korab.

Table D-22: PEGASE European System Test Cases

name	description
case89pegase	89-bus portion of European transmission system from PEGASE project*
case1354pegase	1354-bus portion of European transmission system from PEGASE project*
case2869pegase	2869-bus portion of European transmission system from PEGASE project*
case9241pegase	9241-bus portion of European transmission system from PEGASE project*
case13659pegase	13659-bus portion of European transmission system from PEGASE project*

^{*} Contributed by Cédric Josz and colleagues from the French Transmission System Operator. Please cite references [45, 46] when publishing results based on this data.

Table D-23: RTE French System Test Cases

name	description
case1888rte	1888-bus snapshot of VHV French transmission system from RTE^*
case1951rte	1951-bus snapshot of VHV French transmission system from RTE*
case2848rte	2848-bus snapshot of VHV French transmission system from RTE*
case2868rte	2868-bus snapshot of VHV French transmission system from RTE*
case6468rte	6468-bus snapshot of VHV and HV French transmission system from RTE^*
case6470rte	6470-bus snapshot of VHV and HV French transmission system from RTE^*
case6495rte	6495 -bus snapshot of VHV and HV French transmission system from RTE *
case6515rte	6515-bus snapshot of VHV and HV French transmission system from RTE^*

^{*} Contributed by Cédric Josz and colleagues from the French Transmission System Operator. Please cite reference [45] when publishing results based on this data.

D.4 Automated Test Suite

All of the functions and files listed in this section are found in <MATPOWER>/lib/t, unless noted otherwise.

Table D-24: Automated Test Functions from MP-Test †

name	description	
mptest/lib/		
${ t t}_{ t begin}$	begin running tests	
${ t t_{-}end}$	finish running tests and print statistics	
$t_{ extsf{-}}$ is	tests if two matrices are identical to with a specified tolerance	
t_ok	tests if a condition is true	
t_run_tests	run a series of tests	
t_skip	skips a number of tests, with explanatory message	
t/	MP-Test tests	
test_mptest	runs full MP-Test test suite	
t_test_fcns	$test\ t_ok\ and\ t_is$	

 $^{^\}dagger$ These functions are part of MP-Test and are found in $<\!MATPOWER\!>\!/mptest/lib.$

Table D-25: MIPS $Tests^{\dagger}$

name	description
mips/lib/	
t/	MIPS tests
${ t test_mips}$	runs full MIPS test suite
${ t t}_{ t mips}$	runs tests for MIPS NLP solver
$t_mips_pardiso$	runs tests for MIPS NLP solver, using PARDISO as linear solver
${ t t}$ mplinsolve	tests for mplinsolve
t_qps_mips	runs tests for qps_mips

 $^{^{\}dagger}$ These tests are part of MIPS and are found in $<\!MATPOWER\!>\!/mips/lib/t.$

Table D-26: Test Data

name	description
opf_nle_fcn1.m	user-defined nonlinear equality constraint function for OPF tests
opf_nle_hess1.m	user-defined nonlinear equality constraint Hessian for OPF tests
${\tt pretty_print_acopf.txt}$	pretty-printed output of an example AC OPF run
$pretty_print_dcopf.txt$	pretty-printed output of an example DC OPF run
${\tt soln9_dcopf.mat}$	solution data, DC OPF of t_case9_opf
${\tt soln9_dcpf.mat}$	solution data, DC power flow of t_case9_pf
$soln9_opf_ang.mat$	solution data, AC OPF of t_case9_opfv2 w/only branch angle
	difference limits (gen PQ capability constraints removed)
$soln9_opf_extras1.mat$	solution data, AC OPF of t_case9_opf w/extra cost/constraints
$soln9_opf_Plim.mat$	solution data, AC OPF of t_case9_opf w/opf.flow_lim = 'P'
soln9_opf_PQcap.mat	solution data, AC OPF of t_case9_opfv2 w/only gen PQ capabil-
	ity constraints (branch angle diff limits removed)
$soln9_opf_vg.mat$	solution data, AC OPF of t_case9_opf with option opf.use_vg
$soln9_opf.mat$	solution data, AC OPF of t_case9_opf
$soln9_pf.mat$	solution data, AC power flow of t_case9_pf
$t_{auction_case.m}$	case data used to test auction code
t_case_ext.m	case data used to test ext2int and int2ext, external indexing
t_case_info_eg.txt	example output from case_info, used by t_islands
t_case_int.m	case data used to test ext2int and int2ext, internal indexing
$t_case9_dcline.m$	same as t_case9_opfv2 with additional DC line data
$t_case9_opf.m$	sample case file with OPF data, version 1 format
t_case9_opfv2.m	sample case file with OPF data, version 2 format, includes addi-
	tional branch angle diff limits & gen PQ capability constraints
t_case9_pf.m	sample case file with only power flow data, version 1 format
t_case9_pfv2.m	sample case file with only power flow data, version 2 format
$t_case9_save2psse.m$	sample 9-bus case file used by t_psse for testing save2psse
$t_case9_save2psse.raw$	result of saving t_case9_save2psse.m with save2psse
$t_case30_userfcns.m$	sample case file with OPF, reserves and interface flow limit data
$ t_chgtab.m$	sample change table data, used by t_apply_changes
$t_psse_case.raw$	sample PSS/E format RAW file used to test conversion code
$t_psse_case2.m$	result of converting t_psse_case2.raw to MATPOWER format
$t_psse_case2.raw$	sample PSS/E format RAW file used to test conversion code
$t_psse_case3.m$	result of converting t_psse_case3.raw to MATPOWER format
$t_psse_case3.raw$	sample PSS/E format RAW file used to test conversion code

Table D-27: Miscellaneous Matpower Tests

name	description
test_matpower	runs full Matpower test suite
test_most [‡]	runs full MOST test suite
${ t t_apply_changes}$	runs tests for apply_changes
${ t t_auction_minopf}$	runs tests for auction using MINOPF †
${ t t_auction_mips}$	runs tests for auction using MIPS
$t_auction_tspopf_pdipm$	runs tests for auction using $PDIPM^{\dagger}$
$t_{\tt}ext2int2ext$	runs tests for ext2int and int2ext
$t_feval_w_path$	runs tests for feval_w_path
t_get_losses	runs tests for get_losses
t_hasPQcap	runs tests for hasPQcap
t_have_fcn	runs tests for have_fcn
t_hessian	runs tests for 2 nd derivative code
$\mathtt{t_islands}$	runs test for find_islands and extract_islands
$\mathtt{t}_{ extsf{-}}\mathtt{jacobian}$	runs test for partial derivative code
${ t t_load2disp}$	runs tests for load2disp
${ t t_{-} load case}$	runs tests for loadcase
$t_makeLODF$	runs tests for makeLODF
$t_makePTDF$	runs tests for makePTDF
$t_margcost$	runs tests for margcost
$t_{miqps_matpower}$	runs tests for miqps_matpower
$t_modcost$	runs tests for modcost
$t_{mpoption}$	runs tests for mpoption
$t_nested_struct_copy$	runs tests for nested_struct_copy
$t_{ extsf{o}}$ f2case	runs tests for off2case
${ t t}_{ t opf}{ t model}$	runs tests for opf_model and opt_model objects
${ t t_opf_model_legacy}$	runs tests for opf_model and opt_model objects using legacy (dep-
	recated) add_constraints method
${ t t}_{ t printpf}$	runs tests for printpf
${ t t}_{ t psse}$	runs tests for psse2mpc and related functions
${ t t_qps_matpower}$	runs tests for qps_matpower
$ extsf{t_runmarket}$	runs tests for runmarket
t_scale_load	runs tests for scale_load
t_total_load	runs tests for total_load
t_totcost	runs tests for totcost

 $^{^\}dagger$ Requires the installation of an optional package. See Appendix G for details on the corresponding package.

[‡] While test_most is listed here with other tests, it is actually located in <marpowers/most/lib/t, not <marpowers/lib/t. MOST and its supporting files and functions in the most/ sub-directory are documented in the MOST User's Manual and listed in its Appendix A.

Table D-28: MATPOWER Power Flow, CPF and OPF Tests

name	description	
t_pf	runs tests for AC and DC power flow	
t_pf_radial	runs tests for AC power flow for radial distribution systems	
$t_\mathtt{cpf}$	runs tests for AC continuation power flow	
t_cpf_cb1	example CPF callback function for t_cpf	
t_cpf_cb2	example CPF callback function with cb_args for t_cpf	
$t_opf_dc_bpmpd$	runs tests for DC OPF solver using BPMPD_MEX [†]	
$t_opf_dc_clp$	runs tests for DC OPF solver using CLP [†]	
$t_opf_dc_cplex$	runs tests for DC OPF solver using CPLEX [†]	
$t_opf_dc_glpk$	runs tests for DC OPF solver using GLPK [†]	
$t_opf_dc_gurobi$	runs tests for DC OPF solver using Gurobi [†]	
$t_opf_dc_ipopt$	runs tests for DC OPF solver using $IPOPT^{\dagger}$	
$t_opf_dc_mosek$	runs tests for DC OPF solver using MOSEK [†]	
$t_opf_dc_ot$	runs tests for DC OPF solver using Matlab Opt Toolbox	
$t_opf_dc_mips$	runs tests for DC OPF solver using MIPS	
$t_opf_dc_mips_sc$	runs tests for DC OPF solver using MIPS-sc	
${ t t}_{ t opf}_{ t default}$	runs tests for AC OPF solver using default solver	
${ t t}_{ t o}{ t pf}_{ t f}{ t mincon}$	runs tests for AC OPF solver using fmincon	
${ t t}_{ t o}{ t o}{ t f}_{ t i}{ t p}{ t o}{ t p}{ t t}$	runs tests for AC OPF solver using IPOPT [†]	
${ t t}_{ t opf}$ _knitro	runs tests for AC OPF solver using Artelys Knitro [†]	
$\verb"t_opf_minopf"$	runs tests for AC OPF solver using MINOPF [†]	
${ t t}_{ t o}{ t pf}_{ t m}{ t i}{ t ps}$	runs tests for AC OPF solver using MIPS	
${ t t_opf_mips_sc}$	runs tests for AC OPF solver using MIPS-sc	
$t_{-}opf_{-}softlims$	runs tests for DC OPF with user callback functions for branch flow soft limits	
+		
t_opf_tspopf_pdipm	runs tests for AC OPF solver using PDIPM [†]	
t_opf_tspopf_scpdipm	runs tests for AC OPF solver using SC-PDIPM [†]	
t_opf_tspopf_tralm	runs tests for AC OPF solver using TRALM†	
$t_{\mathtt{opf}}_{\mathtt{userfcns}}$	runs tests for AC OPF with user callback functions for reserves	
+ munonf mod	and interface flow limits	
t_runopf_w_res t dcline	runs tests for AC OPF with fixed reserve requirements	
	runs tests for DC line implementation in toggle_dcline	
t_vdep_load	runs PF, CPF and OPF tests for voltage dependent (ZIP) loads	

 $^{^{\}dagger}$ Requires the installation of an optional package. See Appendix ${\bf G}$ for details on the corresponding package.

Appendix E Matpower Extras

There are a number of MATPOWER-related "extras" that are not part of the core MATPOWER code, but are nevertheless potentially useful to MATPOWER users. Some of these packages and functions were developed and contributed by others. While these extras are not included in the main MATPOWER GitHub repository, they are included in an extras subdirectory in the distribution ZIP files for versioned releases available from the MATPOWER website.

Unless indicated otherwise in the list below, the MATPOWER Extras GitHub repository at https://github.com/MATPOWER/matpower-extras serves as the master repository and should be used for submitting issues or pull requests. Some have their own master repository and are included in MATPOWER Extras as a subrepousing git-subrepo.⁶²

Some of these could be moved into the main MATPOWER distribution in the future with a bit of polishing and additional documentation. Please contact the developers if you are interested in helping make this happen.

maxloadlim

An optimal power flow extension, contributed by Camille Hamon, for computing maximum loadability limits. Please see extras/maxloadlim/manual/maxloadlim_manual.pdf for details on the formulation and implementation. The master repository for this code is at https://github.com/MATPOWER/mx-maxloadlim⁶³ and is included in MATPOWER Extras as a subrepo.

misc

A number of potentially useful functions that are either not yet fully implemented, tested, documented and/or supported. See the help (and the code) in each individual file to understand what it does. 64

reduction

A network reduction toolbox that performs a modified Ward reduction and can be used to produce a smaller approximate equivalent from a larger original system. Code contributed by Yujia Zhu and Daniel Tylavsky. For more details, please see the <code>Network_Reduction_Toolbox.pdf</code> file. The master repository for

⁶²See https://github.com/ingydotnet/git-subrepo.

⁶³This is a fork of Camille Hamon's GitHub repository of this code at:

https://github.com/CamilleH/Max-Load-Lim-matpower

⁶⁴For more information on qcqp_opf, see [45]. For more information on plot_mpc, see [47].

this code is at https://github.com/MATPOWER/mx-reduction and is included in MATPOWER Extras as a subrepo.

sdp_pf

Applications of a semidefinite programming programming relaxation of the power flow equations. Code contributed by Dan Molzahn. See Appendix G.12 and the documentation in the <math align="mailto:modelequation-phistance-phistan

se

State-estimation code contributed by Rui Bo. Type test_se, test_se_14bus or test_se_14bus_err to run some examples. See se_intro.pdf for a brief introduction to this code. The master repository for this code is at https://github.com/MATPOWER/mx-se and is included in MATPOWER Extras as a subrepo.

smartmarket

Code that implements a "smart market" auction clearing mechanism based on MATPOWER's optimal power flow solver. See Appendix F for details.

state_estimator

Older state estimation example, based on code by James S. Thorp.

syngrid

SynGrid is a synthetic grid creation tool for MATPOWER. See the SynGrid User's Manual for details. The master repository for this code is at https://github.com/MATPOWER/mx-syngrid and is included in MATPOWER Extras as a subrepo.

Appendix F "Smart Market" Code

MATPOWER 3 and later includes in the extras/smartmarket directory code that implements a "smart market" auction clearing mechanism. The purpose of this code is to take a set of offers to sell and bids to buy and use MATPOWER's optimal power flow to compute the corresponding allocations and prices. It has been used extensively by the authors with the optional MINOPF package [32] in the context of POWERWEB⁶⁵ but has not been widely tested in other contexts.

The smart market algorithm consists of the following basic steps:

- 1. Convert block offers and bids into corresponding generator capacities and costs.
- 2. Run an optimal power flow with decommitment option (uopf) to find generator allocations and nodal prices (λ_P).
- 3. Convert generator allocations and nodal prices into set of cleared offers and bids.
- 4. Print results.

For step 1, the offers and bids are supplied as two structs, offers and bids, each with fields P for real power and Q for reactive power (optional). Each of these is also a struct with matrix fields qty and prc, where the element in the i-th row and j-th column of qty and prc are the quantity and price, respectively of the j-th block of capacity being offered/bid by the i-th generator. These block offers/bids are converted to the equivalent piecewise linear generator costs and generator capacity limits by the off2case function. See help off2case for more information.

Offer blocks must be in non-decreasing order of price and the offer must correspond to a generator with $0 \le PMIN < PMAX$. A set of price limits can be specified via the lim struct, e.g. and offer price cap on real energy would be stored in lim.P.max_offer. Capacity offered above this price is considered to be withheld from the auction and is not included in the cost function produced. Bids must be in non-increasing order of price and correspond to a generator with PMIN $< PMAX \le 0$ (see Section 6.4.2 on page 70). A lower limit can be set for bids in lim.P.min_bid. See help pricelimits for more information.

The data specified by a MATPOWER case file, with the gen and gencost matrices modified according to step 1, are then used to run an OPF. A decommitment mechanism is used to shut down generators if doing so results in a smaller overall system cost (see Section 8).

⁶⁵See http://www.pserc.cornell.edu/powerweb/.

In step 3 the OPF solution is used to determine for each offer/bid block, how much was cleared and at what price. These values are returned in co and cb, which have the same structure as offers and bids. The mkt parameter is a struct used to specify a number of things about the market, including the type of auction to use, type of OPF (AC or DC) to use and the price limits.

There are two basic types of pricing options available through mkt.auction_type, discriminative pricing and uniform pricing. The various uniform pricing options are best explained in the context of an unconstrained lossless network. In this context, the allocation is identical to what one would get by creating bid and offer stacks and finding the intersection point. The nodal prices (λ_P) computed by the OPF and returned in bus(:,LAM_P) are all equal to the price of the marginal block. This is either the last accepted offer (LAO) or the last accepted bid (LAB), depending which is the marginal block (i.e. the one that is split by intersection of the offer and bid stacks). There is often a gap between the last accepted bid and the last accepted offer. Since any price within this range is acceptable to all buyers and sellers, we end up with a number of options for how to set the price, as listed in Table F-1.

Table F-1: Auction Types

auction type	name	description
0	discriminative	price of each cleared offer (bid) is equal to the offered (bid) price
1	LAO	uniform price equal to the last accepted offer
2	FRO	uniform price equal to the first rejected offer
3	LAB	uniform price equal to the last accepted bid
4	FRB	uniform price equal to the first rejected bid
5	first price	uniform price equal to the offer/bid price of the marginal unit
6	second price	uniform price equal to min(FRO, LAB) if the marginal unit is an offer, or max(FRB, LAO) if it is a bid
7	split-the-difference	uniform price equal to the average of the LAO and LAB
8	dual LAOB	uniform price for sellers equal to LAO, for buyers equal to LAB

Generalizing to a network with possible losses and congestion results in nodal prices λ_P which vary according to location. These λ_P values can be used to normalize all bids and offers to a reference location by multiplying by a locational scale factor. For bids and offers at bus i, this scale factor is $\lambda_P^{\text{ref}}/\lambda_P^i$, where λ_P^{ref} is the nodal price at the reference bus. The desired uniform pricing rule can then be applied to the adjusted offers and bids to get the appropriate uniform price at the reference

bus. This uniform price is then adjusted for location by dividing by the locational scale factor. The appropriate locationally adjusted uniform price is then used for all cleared bids and offers.⁶⁶ The relationships between the OPF results and the pricing rules of the various uniform price auctions are described in detail in [48].

There are certain circumstances under which the price of a cleared offer determined by the above procedures can be less than the original offer price, such as when a generator is dispatched at its minimum generation limit, or greater than the price cap lim.P.max_cleared_offer. For this reason, all cleared offer prices are clipped to be greater than or equal to the offer price but less than or equal to lim.P.max_cleared_offer. Likewise, cleared bid prices are less than or equal to the bid price but greater than or equal to lim.P.min_cleared_bid.

F.1 Handling Supply Shortfall

In single sided markets, in order to handle situations where the offered capacity is insufficient to meet the demand under all of the other constraints, resulting in an infeasible OPF, we introduce the concept of emergency imports. We model an import as a fixed injection together with an equally sized dispatchable load which is bid in at a high price. Under normal circumstances, the two cancel each other and have no effect on the solution. Under supply shortage situations, the dispatchable load is not fully dispatched, resulting in a net injection at the bus, mimicking an import. When used in conjunction with the LAO pricing rule, the marginal load bid will not set the price if all offered capacity can be used.

F.2 Example

The case file lib/t/t_auction_case.m, used for this example, is a modified version of the 30-bus system that has 9 generators, where the last three have negative PMIN to model the dispatchable loads.

- Six generators with three blocks of capacity each, offering as shown in Table F-2.
- Fixed load totaling 151.64 MW.
- Three dispatchable loads, bidding three blocks each as shown in Table F-3.

⁶⁶In versions of MATPOWER prior to 4.0, the smart market code incorrectly shifted prices instead of scaling them, resulting in prices that, while falling within the offer/bid "gap" and therefore acceptable to all participants, did not necessarily correspond to the OPF solution.

Table F-2: Generator Offers

Generator	Block 1 MW @ \$/MWh	Block 2 MW @ \$/MWh	Block 3 MW @ \$/MWh
1	12 @ \$20	24 @ \$50	24 @ \$60
2	12 @ \$20	24 @ \$40	24 @ \$70
3	12 @ \$20	24 @ \$42	24 @ \$80
4	12 @ \$20	24 @ \$44	24 @ \$90
5	12 @ \$20	24 @ \$46	24 @ \$75
6	12 @ \$20	24 @ \$48	24 @ \$60

Table F-3: Load Bids

Load	Block 1 MW @ \$/MWh	Block 2 MW @ \$/MWh	Block 3 MW @ \$/MWh
1	10 @ \$100	10 @ \$70	10 @ \$60
2	10 @ \$100	10 @ \$50	10 @ \$20
3	10 @ \$100	10 @ \$60	10 @ \$50

To solve this case using an AC optimal power flow and a last accepted offer (LAO) pricing rule, we use:

```
mkt.OPF = 'AC';
mkt.auction_type = 1;
```

and set up the problem as follows:

```
mpc = loadcase('t_auction_case');
offers.P.qty = [ ...
        12 24 24;
        12 24 24;
       12 24 24;
        12 24 24;
        12 24 24;
        12 24 24 ];
offers.P.prc = [ ...
        20 50 60;
        20 40 70;
        20 42 80;
        20 44 90;
        20 46 75;
        20 48 60 ];
bids.P.qty = [ ...
        10 10 10;
        10 10 10;
        10 10 10 ];
bids.P.prc = [ ...
        100 70 60;
        100 50 20;
       100 60 50 ];
[mpc_out, co, cb, f, dispatch, success, et] = runmarket(mpc, offers, bids, mkt);
```

The resulting cleared offers and bids are:

```
>> co.P.qty
ans =
                   0
  12.0000 23.3156
  12.0000 24.0000
  12.0000 24.0000 0
12.0000 24.0000 0
12.0000 24.0000 0
12.0000 24.0000 0
>> co.P.prc
ans =
  50.0000 50.0000 50.0000
  50.2406 50.2406 50.2406
  50.3368 50.3368 50.3368
  51.0242 51.0242 51.0242
  52.1697 52.1697 52.1697
  52.9832 52.9832 52.9832
>> cb.P.qty
ans =
  10.0000 10.0000 10.0000
  10.0000 0 0
  10.0000 10.0000 0
>> cb.P.prc
ans =
  51.8207 51.8207 51.8207
  54.0312 54.0312 54.0312
  55.6208 55.6208
                    55.6208
```

In other words, the sales by generators and purchases by loads are as shown summarized in Tables F-4 and Tables F-5, respectively.

Table F-4: Generator Sales

Generator	Quantity Sold MW	Selling Price \$/MWh
1	35.3	\$50.00
2	36.0	\$50.24
3	36.0	\$50.34
4	36.0	\$51.02
5	36.0	\$52.17
6	36.0	\$52.98

Table F-5: Load Purchases

Load	Quantity Bought MW	Purchase Price \$/MWh
1	30.0	\$51.82
2	10.0	\$54.03
3	20.0	\$55.62

F.3 Smartmarket Files and Functions

Table F-6: Smartmarket Files and Functions

name	description
extras/smartmarket/	
auction	clears set of bids and offers based on pricing rules and OPF results
case2off	generates quantity/price offers and bids from gen and gencost
$\mathtt{idx_disp}$	named column index definitions for dispatch matrix
off2case	updates gen and gencost based on quantity/price offers and bids
pricelimits	fills in a struct with default values for offer and bid limits
printmkt	prints the market output
runmarket	top-level simulation function, runs the OPF-based smart market
\mathtt{runmkt}^*	top-level simulation function, runs the OPF-based smart market
smartmkt	implements the smart market solver
SM_CHANGES	change history for the smart market software

 $^{^{*}}$ Deprecated. Will be removed in a subsequent version.

Appendix G Optional Packages

There are a number of optional packages, not included in the MATPOWER distribution, that MATPOWER can utilize if they are installed in your MATLAB path. Each of them is based on one or more MEX files pre-compiled for various platforms, some distributed by PSERC, others available from third parties, and each with their own terms of use.

G.1 BPMPD_MEX – MEX interface for BPMPD

BPMPD_MEX [30, 31] is a MATLAB MEX interface to BPMPD, an interior point solver for quadratic programming developed by Csaba Mészáros at the MTA SZ-TAKI, Computer and Automation Research Institute, Hungarian Academy of Sciences, Budapest, Hungary. It can be used by MATPOWER'S DC and LP-based OPF solvers and it improves the robustness of MINOPF. It is also useful outside of MATPOWER as a general-purpose QP/LP solver.

This MEX interface for BPMPD was coded by Carlos E. Murillo-Sánchez, while he was at Cornell University. It does not provide all of the functionality of BPMPD, however. In particular, the stand-alone BPMPD program is designed to read and write results and data from MPS and QPS format files, but this MEX version does not implement reading data from these files into MATLAB.

The current version of the MEX interface is based on version 2.21 of the BPMPD solver, implemented in Fortran.

Builds are available for Linux (32-bit), Mac OS X (PPC, Intel 32-bit) and Windows (32-bit) at http://www.pserc.cornell.edu/bpmpd/.

When installed BPMPD_MEX can be selected as the solver for DC OPFs by setting the opf.dc.solver option to 'BPMPD'. It can also be used to solve general LP and QP problems via MATPOWER's common QP solver interface qps_matpower with the algorithm option set to 'BPMPD' (or 100 for backward compatibility), or by calling qps_bpmpd directly.

G.2 CLP - COIN-OR Linear Programming

The CLP [37] (COIN-OR Linear Programming) solver is an open-source linear programming solver written in C++ by John Forrest. It can solve both linear programming (LP) and quadratic programming (QP) problems. It is primarily meant to be used as a callable library, but a basic, stand-alone executable version exists as well. It is available from the COIN-OR initiative at https://github.com/coin-or/Clp.

To use CLP with MATPOWER, a MEX interface is required⁶⁷. For Microsoft Windows users, a pre-compiled MEX version of CLP (and numerous other solvers, such as GLPK and IPOPT) are easily installable as part of the OPTI Toolbox⁶⁸ [49]

With the Matlab interface to CLP installed, it can be used to solve DC OPF problems by setting the opf.dc.solver option equal to 'CLP'. The solution algorithms and other CLP parameters can be set directly via Matpower's clp.opts option. A "CLP user options" function can also be specified via clp.opt_fname to override the defaults for any of the many CLP parameters. See help clp for details. See Table C-8 for a summary of the CLP-related Matpower options.

CLP can also be used to solve general LP and QP problems via MATPOWER'S common QP solver interface qps_matpower with the algorithm option set to 'CLP', or by calling qps_clp directly.

G.3 CPLEX – High-performance LP and QP Solvers

The IBM ILOG CPLEX Optimizer, or simply CPLEX, is a collection of optimization tools that includes high-performance solvers for large-scale linear programming (LP) and quadratic programming (QP) problems, among others. More information is available at https://www.ibm.com/analytics/cplex-optimizer.

Although CPLEX is a commercial package, at the time of this writing the full version is available to academics at no charge through the IBM Academic Initiative program for teaching and non-commercial research. See http://www.ibm.com/support/docview.wss?uid=swg21419058 for more details.

When the Matlab interface to CPLEX is installed, the CPLEX LP and QP solvers, cplexlp and cplexqp, can be used to solve DC OPF problems by setting the opf.dc.solver option equal to 'CPLEX'. The solution algorithms can be controlled by Matpower's cplex.lpmethod and cplex.qpmethod options. See Table C-9 for a summary of the CPLEX-related Matpower options. A "CPLEX user options" function can also be specified via cplex.opt_fname to override the defaults for any of the many CPLEX parameters. See help cplex_options and the "Parameters of

⁶⁷According to David Gleich at http://web.stanford.edu/~dgleich/notebook/2009/03/coinor_clop_for_matlab.html, there was a MATLAB MEX interface to CLP written by Johan Lofberg and available (at some point in the past) at http://control.ee.ethz.ch/~joloef/mexclp.zip. Unfortunately, at the time of this writing, it seems it is no longer available there, but Davide Barcelli makes some precompiled MEX files for some platforms available here http://www.dii.unisi.it/~barcelli/software.php, and the ZIP file linked as Clp 1.14.3 contains the MEX source as well as a clp.m wrapper function with some rudimentary documentation.

68The OPTI Toolbox is available from https://www.inverseproblem.co.nz/OPTI/.

CPLEX" section of the CPLEX documentation at http://www.ibm.com/support/knowledgecenter/SSSA5P for details.

It can also be used to solve general LP and QP problems via MATPOWER's common QP solver interface qps_matpower, or MILP and MIQP problems via miqps_matpower, with the algorithm option set to 'CPLEX' (or 500 for backward compatibility), or by calling qps_cplex or miqps_cplex directly.

G.4 GLPK – GNU Linear Programming Kit

The GLPK [36] (**G**NU **L**inear **P**rogramming **K**it) package is intended for solving large-scale linear programming (LP), mixed integer programming (MIP), and other related problems. It is a set of routines written in ANSI C and organized in the form of a callable library.

To use GLPK with MATPOWER, a MEX interface is required⁶⁹. For Microsoft Windows users, a pre-compiled MEX version of GLPK (and numerous other solvers, such as CLP and IPOPT) are easily installable as part of the OPTI Toolbox⁷⁰ [49].

When GLPK is installed, either as part of Octave or with a MEX interface for MATLAB, it can be used to solve DC OPF problems that do not include any quadratic costs by setting the opf.dc.solver option equal to 'GLPK'. The solution algorithms and other GLPK parameters can be set directly via MATPOWER's glpk.opts option. A "GLPK user options" function can also be specified via glpk.opt_fname to override the defaults for any of the many GLPK parameters. See help glpk_options and the parameters section the GLPK documentation at https://www.gnu.org/software/octave/doc/interpreter/Linear-Programming.html for details. See Table C-11 for a summary of the GLPK-related MATPOWER options.

GLPK can also be used to solve general LP problems via MATPOWER's common QP solver interface qps_matpower, or MILP problems via miqps_matpower, with the algorithm option set to 'GLPK', or by calling qps_glpk or miqps_glpk directly.

G.5 Gurobi – High-performance LP and QP Solvers

Gurobi [35] is a collection of optimization tools that includes high-performance solvers for large-scale linear programming (LP) and quadratic programming (QP) problems,

⁶⁹The http://glpkmex.sourceforge.net site and Davide Barcelli's page http://www.dii.unisi.it/~barcelli/software.php may be useful in obtaining the MEX source or pre-compiled binaries for Mac or Linux platforms.

⁷⁰The OPTI Toolbox is available from https://www.inverseproblem.co.nz/OPTI/.

among others. The project was started by some former CPLEX developers. More information is available at https://www.gurobi.com/.

Although Gurobi is a commercial package, at the time of this writing their is a free academic license available. See https://www.gurobi.com/academia/for-universities for more details.

Beginning with version 5.0.0, Gurobi includes a native MATLAB interface, which is supported in MATPOWER version 4.2 and above.⁷¹

When Gurobi is installed, it can be used to solve DC OPF problems by setting the opf.dc.solver option equal to 'GUROBI'. The solution algorithms can be controlled by MATPOWER's gurobi.method option. See Table C-12 for a summary of the Gurobi-related MATPOWER options. A "Gurobi user options" function can also be specified via gurobi.opt_fname to override the defaults for any of the many Gurobi parameters. See help gurobi_options and the "Parameters" section of the "Gurobi Optimizer Reference Manual" at https://www.gurobi.com/documentation/ for details.

It can also be used to solve general LP and QP problems via MATPOWER's common QP solver interface qps_matpower, or MILP and MIQP problems via miqps_matpower, with the algorithm option set to 'GUROBI' (or 700 for backward compatibility), or by calling qps_gurobi or miqps_gurobi directly.

G.6 IPOPT - Interior Point Optimizer

IPOPT [51] (Interior Point **OPT**imizer, pronounced I-P-Opt) is a software package for large-scale nonlinear optimization. It is is written in C++ and is released as open source code under the Common Public License (CPL). It is available from the COIN-OR initiative at https://github.com/coin-or/Ipopt. The code has been written by Carl Laird and Andreas Wächter, who is the COIN project leader for IPOPT.

MATPOWER requires the MATLAB MEX interface to IPOPT, which is included in the IPOPT source distribution, but must be built separately. Additional information on the MEX interface is available at https://projects.coin-or.org/Ipopt/wiki/MatlabInterface. Please consult the IPOPT documentation, web-site and mailing lists for help in building and installing the IPOPT MATLAB interface. This interface uses callbacks to MATLAB functions to evaluate the objective function and its gradient, the constraint values and Jacobian, and the Hessian of the Lagrangian.

⁷¹MATPOWER version 4.1 supported Gurobi version 4.x, which required a free third-party MATLAB MEX interface [50], available at http://www.convexoptimization.com/wikimization/index.php/Gurobi_mex.

Precompiled MEX binaries for a high-performance version of IPOPT, using the PARDISO linear solver [52,53], are available from the PARDISO project⁷². At the time of this writing, these are the highest performing solvers available to MATPOWER for very large scale AC OPF problems. For Microsoft Windows users, a pre-compiled MEX version of IPOPT (and numerous other solvers, such as CLP and GLPK) are easily installable as part of the OPTI Toolbox⁷³ [49].

When installed, IPOPT can be used by MATPOWER to solve both AC and DC OPF problems by setting the opf.ac.solver or opf.dc.solver options, respectively, equal to 'IPOPT'. See Table C-13 for a summary of the IPOPT-related MATPOWER options. The many algorithm options can be set by creating an "IPOPT user options" function and specifying it via ipopt.opt_fname to override the defaults set by ipopt_options. See help ipopt_options and the options reference section of the IPOPT documentation at http://www.coin-or.org/Ipopt/documentation/ for details.

It can also be used to solve general LP and QP problems via MATPOWER's common QP solver interface qps_matpower with the algorithm option set to 'IPOPT' (or 400 for backward compatibility), or by calling qps_ipopt directly.

G.7 Artelys Knitro – Non-Linear Programming Solver

Artelys Knitro [34] is a general purpose optimization solver specializing in nonlinear problems, available from Artelys. As of version 9, Knitro includes a native MATLAB interface, knitromatlab⁷⁴. More information is available at https://www.artelys.com/solvers/knitro/ and https://www.artelys.com/docs/knitro/.

Although Artelys Knitro is a commercial package, at the time of this writing there is a free academic license available, with details on their download page.

When installed, Knitro's MATLAB interface function, knitromatlab or ktrlink, can be used by MATPOWER to solve AC OPF problems by simply setting the opf.ac.solver option to 'KNITRO'. See Table C-14 for a summary of Knitro-related MATPOWER options. The knitromatlab function uses callbacks to MATLAB functions to evaluate the objective function and its gradient, the constraint values and Jacobian, and the Hessian of the Lagrangian.

Knitro options can be controlled directly by creating a standard Knitro options

⁷²See https://pardiso-project.org/ for the download links.

⁷³The OPTI Toolbox is available from https://www.inverseproblem.co.nz/OPTI/.

⁷⁴Earlier versions required the MATLAB Optimization Toolbox from The MathWorks, which included an interface to the Knitro libraries called ktrlink, but the libraries themselves still had to be acquired directly from Ziena Optimization, LLC (subsequently acquired by Artelys).

file in your working directory and specifying it via the knitro.opt_fname (or, for backward compatibility, naming it knitro_user_options_n.txt and setting MATPOWER's knitro.opt option to n, where n is some positive integer value). See the Artelys Knitro user manuals at https://www.artelys.com/docs/knitro/ for details on the available options.

G.8 MINOPF – AC OPF Solver Based on MINOS

MINOPF [32] is a MINOS-based optimal power flow solver for use with MATPOWER. It is for educational and research use only. MINOS [33] is a legacy Fortran-based software package, developed at the Systems Optimization Laboratory at Stanford University, for solving large-scale optimization problems.

While MINOPF is often MATPOWER's fastest AC OPF solver on small problems, as of MATPOWER 4, it no longer becomes the default AC OPF solver when it is installed. It can be selected manually by setting the opf.ac.solver option to 'MINOPF' (see help mpoption for details).

Builds are available for Linux (32-bit), Mac OS X (PPC, Intel 32-bit) and Windows (32-bit) at http://www.pserc.cornell.edu/minopf/.

G.9 MOSEK – High-performance LP and QP Solvers

MOSEK is a collection of optimization tools that includes high-performance solvers for large-scale linear programming (LP) and quadratic programming (QP) problems, among others. More information is available at https://www.mosek.com/.

Although MOSEK is a commercial package, at the time of this writing there is a free academic license available. See https://www.mosek.com/products/academic-licenses/for more details.

When the MATLAB interface to MOSEK is installed, the MOSEK LP and QP solvers can be used to solve DC OPF problems by setting the opf.dc.solver option equal to 'MOSEK'. The solution algorithm for LP problems can be controlled by MATPOWER's mosek.lp_alg option. See Table C-16 for other MOSEK-related MATPOWER options. A "MOSEK user options" function can also be specified via mosek.opt_fname to override the defaults for any of the many MOSEK parameters. For details see help mosek_options and the "Parameters" reference in "The MOSEK optimization toolbox for MATLAB manual" at http://docs.mosek.com/7.1/toolbox/Parameters.html. You may also find it helpful to use the symbolic constants defined by mosek_symboon.

It can also be used to solve general LP and QP problems via MATPOWER's common QP solver interface qps_matpower, or MILP and MIQP problems via miqps_matpower, with the algorithm option set to 'MOSEK' (or 600 for backward compatibility), or by calling qps_mosek or miqps_mosek directly.

G.10 Optimization Toolbox – LP, QP, NLP and MILP Solvers

MATLAB's Optimization Toolbox [29,54], available from The MathWorks, provides a number of high-performance solvers that MATPOWER can take advantage of.

It includes fmincon for nonlinear programming problems (NLP), and linprog and quadprog for linear programming (LP) and quadratic programming (QP) problems, respectively. For mixed-integer linear programs (MILP), it provides intlingprog. Each solver implements a number of different solution algorithms. More information is available from The MathWorks, Inc. at https://www.mathworks.com/.

When available, the Optimization Toolbox solvers can be used to solve AC or DC OPF problems by setting the opf.ac.solver or opf.dc.solver options, respectively, equal to 'OT'.

The solution algorithm used by fmincon for NLP problems can be controlled by MATPOWER's fmincon.alg option. See Table C-10 for other MATPOWER options related to fmincon. For linprog, quadprog and intlingprog, the corresponding MATPOWER option can be used to pass in native Optimization Toolbox options directly to the solver. For example, to set the LP solver to use a dual simplex method, simply set MATPOWER's 'linprog.Algorithm' option to 'dual-simplex'. For details on the full set of Optimization Toolbox options, please refer to their documentation.⁷⁵

The Optimization Toolbox can also be used to solve general LP and QP problems via MATPOWER's common QP solver interface qps_matpower, or MILP problems via miqps_matpower, with the algorithm option set to 'OT', or by calling qps_ot or miqps_ot directly.

G.11 PARDISO – Parallel Sparse Direct and Multi-Recursive Iterative Linear Solvers

The PARDISO package is a thread-safe, high-performance, robust, memory efficient and easy to use software for solving large sparse symmetric and non-symmetric linear systems of equations on shared-memory and distributed-memory multiprocessor systems [52,53]. More information is available at https://pardiso-project.org.

⁷⁵See https://www.mathworks.com/help/optim/ and [54].

When the MATLAB interface to PARDISO is installed, PARDISO's solvers can be used to replace the built-in \setminus operator for solving for the Newton update step in MATPOWER's default primal-dual interior point solver (MIPS) by setting the mips.linsolver option equal to 'PARDISO'. The mplinsolve function can also be called directly to solve Ax = b problems via PARDISO or the built-in solver, depending on the arguments supplied. This interface also gives access to the full range of PARDISO's options. For details, see help mplinsolve and the PARDISO User's Manual at https://pardiso-project.org/manual/manual.pdf.

When solving very large AC optimal power flow problems with MIPS, selecting PARDISO as the linear solver can often dramtically improve both computation time and memory use.

Also note that precompiled MEX binaries for a high-performance version of IPOPT, using the PARDISO linear solver, are available. Refer to Section G.6 for more details.

G.12 SDP_PF – Applications of a Semidefinite Programming Relaxation of the Power Flow Equations

This package is contributed by Dan Molzahn and is currently included with MAT-POWER in the <mathcal{MATPOWER}/extras/sdp_pf directory. Complete documentation is available in <mathcal{MATPOWER}/extras/sdp_pf/documentation directory, and especially in the file sdp_pf_documentation.pdf.

G.13 TSPOPF – Three AC OPF Solvers by H. Wang

TSPOPF [28] is a collection of three high performance AC optimal power flow solvers for use with MATPOWER. The three solvers are:

- PDIPM primal/dual interior point method
- SCPDIPM step-controlled primal/dual interior point method
- TRALM trust region based augmented Lagrangian method

The algorithms are described in [8,43]. The first two are essentially C-language implementations of the algorithms used by MIPS (see Appendix A), with the exception that the step-controlled version in TSPOPF also includes a cost smoothing technique in place of the constrained-cost variable (CCV) approach for handling piece-wise linear costs.

The PDIPM in particular is significantly faster for large systems than any previous MATPOWER AC OPF solver, including MINOPF. When TSPOPF is installed, the PDIPM solver becomes the default optimal power flow solver for MATPOWER. Additional options for TSPOPF can be set using mpoption (see help mpoption for details).

Builds are available for Linux (32-bit, 64-bit), Mac OS X (PPC, Intel 32-bit, Intel 64-bit) and Windows (32-bit, 64-bit) at http://www.pserc.cornell.edu/tspopf/.

Appendix H Release History

The full release history can be found in docs/CHANGES or online at https://github.com/MATPOWER/matpower/blob/master/CHANGES.md.

H.1 Pre 1.0 – released Jun 25, 1997

Work on MATPOWER began in 1996, with an optimal power flow based on MATLAB's constr function.

H.2 Version 1.0 – released Sep 17, 1997

The MATPOWER 1.0 User's Manual is available online. ⁷⁶

New Features

- Successive LP-based optimal power flow solver, by Deqiang Gan.
- Automatic conversion between possibly non-consecutive external bus numbers and consecutive internal bus numbers.
- IEEE CDF to Matpower case conversion script, by Deqiang Gan.

Other Changes

- Top-level scripts converted to functions that take case file as input.
- Updated case file format. Generator costs moved to separate gencost table.

H.3 Version 1.0.1 – released Sep 19, 1997

Changes

• Bug fixes.

⁷⁶https://matpower.org/docs/MATPOWER-manual-1.pdf

H.4 Version 2.0 – released Dec 24, 1997

The MATPOWER 2.0 User's Manual is available online. 77

New Features

- greatly enhanced output options
- fast-decoupled power flow
- optional costs for reactive power generation in OPF
- consolidated parameters in Matpower options vector

Other Changes

- \bullet optimized building of $Y_{\rm bus}$ and Jacobian matrices (much faster on large systems, especially in Matlab 5)
- highly improved unit decommitment algorithm
- completely rewritten LP-based OPF solver
- various bug fixes

⁷⁷https://matpower.org/docs/MATPOWER-manual-2.0.pdf

H.5 Version 3.0 – released Feb 14, 2005

The Matpower 3.0 User's Manual is available online.⁷⁸

New Features

- Compatibility with Matlab 7 and Optimization Toolbox 3.
- DC power flow and DC OPF solvers added.
- Option to enforce generator reactive power limits in AC power flow solution.
- Gauss-Seidel power flow solver added.
- Support for MINOS-based OPF solver added (separate package, see http://www.pserc.cornell.edu/minopf/ for more details)
- Multiple generators at a single bus.
- Saving of solved cases as M-files or MAT-files.
- Loading of input data from M-files, MAT-files, or structs.
- Improved decommitment algorithm.
- Added a very incomplete test suite.
- Handling of dispatchable loads in OPF, modeled as negative generators with constant power factor constraint.

Bugs Fixed

- Phase shifters shifted the wrong direction.
- Minor fixes to IEEE CDF to Matpower format conversion (reported by D. Devaraj and Venkat)
- Flows on out-of-service lines were not being zeroed out. (reported by Ramazan Caglar)
- Reported total inter-tie flow values and area export values were incorrect.
- Several other bugs in solution printouts.

⁷⁸https://matpower.org/docs/MATPOWER-manual-3.0.pdf

H.6 Version 3.2 – released Sep 21, 2007

The Matpower 3.2 User's Manual is available online.⁷⁹

- AC OPF formulation enhancements
 - new generalized cost model
 - piece-wise linear generator PQ capability curves
 - branch angle difference constraints
 - simplified interface for specifying additional linear constraints
 - option to use current magnitude for line flow limits (set OPF_FLOW_LIM to 2, fmincopf solver only)
- AC OPF solvers
 - support for TSPOPF, a new optional package of three OPF solvers, implemented as C MEX files, suitable for large scale systems
 - ability to specify initial value and bounds on user variables z
- New (v. 2) case file format
 - all data in a single struct
 - generator PQ capability curves
 - generator ramp rates
 - branch angle difference limits
- New function makePTDF to build DC PDTF matrix
- Added 5 larger scale (> 2000 bus) cases for Polish system (thanks to Roman Korab).
- Improved identification of binding constraints in printout.
- Many new tests in test suite.

⁷⁹https://matpower.org/docs/MATPOWER-manual-3.2.pdf

Bugs Fixed

- Phase shifters shifted the wrong direction, again (v.2 had it right).
- Fixed bug in pfsoln which caused incorrect value for reactive generation when $Q_{\min} = Q_{\max}$ for all generators at a bus in power flow solution.

Incompatible Changes

- User supplied A matrix for general linear constraints in OPF no longer includes columns for helper variables for piecewise linear gen costs, and now requires columns for all x (OPF) variables.
- Changed the sign convention used for phase shifters to be consistent with PTI, PowerWorld, PSAT, etc. E.g. A phase shift of 10 deg now means the voltage at the "to" end is delayed by 10 degrees.
- Name of option 24 in mpoption changed from OPF_P_LINE_LIM to OPF_FLOW_LIM.

H.7 Version 4.0 – released Feb 7, 2011

The MATPOWER 4.0 User's Manual is available online.⁸⁰

- Licensed under the GNU General Public License (GPL).
- Added compatibility with GNU Octave, a free, open-source Matlab clone.
- Extensive OPF enhancements:
 - Generalized, extensible OPF formulation applies to all solvers (AC and DC).
 - Improved method for modifying OPF formulation and output via a new user-defined callback function mechanism.
 - Option to co-optimize reserves based on fixed zonal reserve requirements, implemented using new callback function mechanism.
 - Option to include interface flow limits (based on DC model flows), implemented using new callback function mechanism.
- New high performance OPF solvers:
 - MIPS (Matpower Interior Point Solver), a new a pure-Matlab implementation of the primal-dual interior point methods from the optional package TSPOPF. MIPS is suitable for large systems and is used as Matpower's default solver for AC and DC OPF problems if no other optional solvers are installed. To select MIPS explicitly, use OPF_ALG = 560/565 and OPF_ALG_DC = 200/250 for AC and DC OPF, respectively. MIPS can also be used independently of Matpower as a solver for general nonlinear constrained optimization problems.
 - Support for the IPOPT interior point optimizer for large scale nonlinear optimization. Use OPF_ALG = 580 and OPF_ALG_DC = 400 for AC and DC OPF, respectively. Requires the MATLAB MEX interface for IPOPT, available from https://projects.coin-or.org/Ipopt/.
 - Support for CPLEX to solve LP and QP problems. Set option OPF_ALG_DC = 500 to use CPLEX to solve the DC OPF. Requires the MATLAB interface to

⁸⁰https://matpower.org/docs/MATPOWER-manual-4.0.pdf

- CPLEX, available from https://www.ibm.com/analytics/cplex-optimizer. See help mpoption for more CPLEX options.
- Support for MOSEK to solve LP and QP problems. Set option OPF_ALG_DC = 600 to use MOSEK to solve the DC OPF. Requires the MATLAB interface to MOSEK, available from https://www.mosek.com/. See help mpoption for more MOSEK options.
- Updated support for MathWorks' Optimization Toolbox solvers, fmincon(),
 linprog() and quadprog().

• Improved documentation:

- New, rewritten User's Manual (docs/manual.pdf).
- Two new Tech Notes, available from MATPOWER home page.
 - * Uniform Price Auctions and Optimal Power Flow
 - * AC Power Flows, Generalized OPF Costs and their Derivatives using Complex Matrix Notation
- Help text updates to more closely match MathWorks conventions.

• New functions:

- load2disp() converts from fixed to dispatchable loads.
- makeJac() forms the power flow Jacobian. Optionally returns the system admittance matrices too.
- makeLODF() computes line outage distribution factors.
- modcost() shifts/scales generator cost functions.
- qps_matpower() provides a consistent, unified interface to all of MAT-POWER's available QP/LP solvers, serving as a single wrapper around qps_bpmpd(), qps_cplex(), qps_ipopt(), qps_mips(), and qps_ot() (Optimization Toolbox, i.e. quadprog(), linprog()).
- scale_load() conveniently modifies multiple loads.
- total_load() retreives total load for the entire system, a specific zone or bus, with options to include fixed loads, dispatchable loads or both.
- Option to return full power flow or OPF solution in a single results struct, which is a superset of the input case struct.

- Ability to read and save generalized OPF user constraints, costs and variable limits as well as other user data in case struct.
- Numerous performance optimizations for large scale systems.
- Perl script psse2matpower for converting PSS/E data files to MATPOWER case format.
- Deprecated areas data matrix (was not being used).
- Many new tests in test suite.

Bugs Fixed

- Auction code in extras/smartmarket in all previous versions contained a design error which has been fixed. Prices are now scaled instead of shifted when modified according to specified pricing rule (e.g. LAO, FRO, LAB, FRB, splitthe-difference, etc.). Auctions with both real and reactive offers/bids must be type 0 or 5, type 1 = LAO is no longer allowed.
- Branch power flow limits could be violated when using the option OPF_FLOW_LIM = 1.

Incompatible Changes

• Renamed functions used to compute AC OPF cost, constraints and hessian, since they are used by more than fmincon:

```
\begin{array}{l} {\tt costfmin} \to {\tt opf\_costfcn} \\ {\tt consfmin} \to {\tt opf\_consfcn} \\ {\tt hessfmin} \to {\tt opf\_hessfcn} \end{array}
```

- Input/output arguments to uopf() are now consistent with opf().
- dAbr_dV() now gives partial derivatives of the **squared** magnitudes of flows w.r.t. V, as opposed to the magnitudes.

H.8 Version 4.1 – released Dec 14, 2011

The MATPOWER 4.1 User's Manual is available online.⁸¹

New Features

- More new high performance OPF solvers:
 - Support for the KNITRO interior point optimizer for large scale nonlinear optimization. Use OPF_ALG = 600 for to select KNITRO to solve the AC OPF. Requires the MATLAB Optimization Toolbox and a license for KNITRO, available from http://www.ziena.com/.82 See help mpoption for more KNITRO options.
 - Support for Gurobi to solve LP and QP problems. Set option OPF_ALG_DC = 700 to use Gurobi to solve the DC OPF. Requires Gurobi (https://www.gurobi.com/) and the Gurobi MEX interface (http://www.convexoptimization.com/wikimization/index.php/Gurobi_mex). See help mpoption for more Gurobi options.
 - Updated for compatibility with CPLEX 12.3.
 - Changed options so that fmincon uses its interior-point solver by default.
 Much faster on larger systems.
- Support for basic modeling of DC transmission lines.
- New case files with more recent versions of Polish system.
- Power flow can handle networks with islands.

Bugs Fixed

- Computation of quadratic user-defined costs had a potentially fatal error. Thanks to Stefanos Delikaraoglou for finding this.
- Calculation of reserve prices in toggle_reserves() had an error.

Incompatible Changes

• Optional packages TSPOPF and MINOPF must be updated to latest versions.

⁸¹https://matpower.org/docs/MATPOWER-manual-4.1.pdf

⁸² Subsequently acquired by Artelys, http://www.artelys.com/.

H.9 Version 5.0 – released Dec 17, 2014

The Matpower 5.0 User's Manual is available online.⁸³

- Continuation power flow with tangent predictor and Newton method corrector, based on code contributed by Shrirang Abhyankar and Alex Flueck.
- SDP_PF, a set of applications of a semidefinite programming relaxation of the power flow equations, contributed by Dan Molzahn (see extras/sdp_pf):
 - Globally optimal AC OPF solver (under certain conditions).
 - Functions to check sufficient conditions for:
 - * global optimality of OPF solution
 - * insolvability of power flow equations
- PSS/E RAW data conversion to Matpower case format (experimental) based on code contributed by Yujia Zhu.
- Brand new extensible MATPOWER options architecture based on options struct instead of options vector.
- Utility routines to check network connectivity and handle islands and isolated buses.
- New extension implementing DC OPF branch flow soft limits. See help toggle_softlims for details.
- New and updated support for 3rd party solvers:
 - CPLEX 12.6
 - GLPK
 - Gurobi 5.x
 - Ірорт 3.11.х
 - Knitro 9.x.x
 - Optimization Toolbox 7.1

⁸³https://matpower.org/docs/MATPOWER-manual-5.0.pdf

• Numerous performance enhancements.

• New functions:

- runcpf() for continuation power flow.
- case_info() for summarizing system information, including network connectivity.
- extract_islands() to extract a network island into a separate MAT-POWER case.
- find_islands() to detect network islands.
- Copt_model/describe_idx() to identify variable, constraint or cost row indices to aid in debugging.
- margcost() for computing the marginal cost of generation.
- psse2mpc() to convert PSS/E RAW data into MATPOWER case format.
- get_losses() to compute branch series losses and reactive charging injections and derivatives as functions of bus voltages.
- New experimental functions in extras/misc for computing loss factors, checking feasibility of solutions, converting losses to negative bus injections and modifying an OPF problem to make it feasible.
- Added case5.m, a 5-bus, 5-generator example case from Rui Bo.

• New options:

- scale_load() can scale corresponding gencost for dispatchable loads.
- makeJac() can return full Jacobian instead of reduced version used in Newton power flow updates.
- modcost() can accept a vector of shift/scale factors.
- total_load() can return actual or nominal values for dispatchable loads.
- runpf(), runopf(), etc. can send pretty-printed output to file without also sending it to the screen.
- out.suppress_detail option suppresses all output except system summary (on by default for large cases).
- opf.init_from_mpc option forces some solvers to use user-supplied starting point.

- MIPS 1.1 includes many new user-settable options.
- Reimplementated <code>copf_model</code> class as sub-class of the new <code>copt_model</code> class, which supports indexed named sets of variables, constraints and costs.
- Many new tests in test suite.

Bugs Fixed

- Running a power flow for a case with DC lines but no gencost no longer causes an error.
- Fixed a bug in runpf() where it was using the wrong initial voltage magnitude for generator buses marked as PQ buses. Power flow of solved case was not converging in zero iterations as expected.
- Fixed fatal bug in MIPS for unconstrained, scalar problems. Thanks to Han Na Gwon.
- Fixed a bug in int2ext() where converting a case to internal ordering before calling runpf() or runopf() could result in a fatal error due to mismatched number of columns in internal and external versions of data matrices. Thanks to Nasiruzzaman and Shiyang Li for reporting and detailing the issue.
- DC OPF now correctly sets voltage magnitudes to 1 p.u. in results.
- Fixed a bug in MIPS where a near-singular matrix could produce an extremely large Newton step, resulting in incorrectly satisfying the relative feasibility criterion for successful termination.
- Improved the starting point created for IPOPT, Knitro and MIPS for variables that are only bounded on one side.
- Fixed bug in savecase() where the function name mistakenly included the path when the fname input included a path.
- Fixed bugs in runpf() related to enforcing generator reactive power limits when all generators violate limits or when the slack bus is converted to PQ.
- Fixed crash when using Knitro to solve cases with all lines unconstrained.

- Fixed memory issue resulting from nested om fields when repeatedly running an OPF using the results of a previous OPF as input. Thanks to Carlos Murillo-Sánchez.
- Fixed fatal error when uopf() shuts down all gens attempting to satisfy P_{min} limits.
- Reactive power output of multiple generators at a PQ bus no longer get reallocated when running a power flow.
- Fixed a bug in savecase() where a gencost matrix with extra columns of zeros resulted in a corrupted MATPOWER case file.
- Fixed bug in runpf() that caused a crash for cases with pf.enforce_q_lims turned on and exactly two Q limit violations, one Qmax and one Qmin. Thanks to Jose Luis Marín.

Incompatible Changes

- Optional packages TSPOPF and MINOPF must be updated to latest versions.
- Renamed cdf2matp() to cdf2mpc() and updated the interface to be consistent with psse2mpc().
- Removed ot_opts field, replaced with linprog_opts and quadprog_opts fields in the opt argument to qps_matpower() and qps_ot().
- The name of the mips() option used to specify the maximum number of stepsize reductions with step_control on was changed from max_red to sc.red_it for consistency with other MATPOWER options.
- Removed max_it option from qps_matpower() (and friends) args. Use algorithm specific options to control iteration limits.
- Changed behavior of branch angle difference limits so that 0 is interpreted as unbounded only if both ANGMIN and ANGMAX are zero.
- In results struct returned by an OPF, the value of results.raw.output.alg is now a string, not an old-style numeric alg code.

• Removed:

- Support for Matlab 6.x.
- Support for constr() and successive LP-based OPF solvers.
- Support for Gurobi 4.x/gurobi mex() interface.
- extras/cpf, replaced by runcpf().
- extras/psse2matpower, replaced by psse2mpc().

H.10 Version 5.1 – released Mar 20, 2015

The MATPOWER 5.1 User's Manual is available online.⁸⁴

New License

• Switched to the more permissive 3-clause BSD license from the previously used GNU General Public License (GPL) v3.0.

New Case Files

• Added four new case files, ranging from 89 up to 9421 buses, representing parts of the European high voltage transmission network, stemming from the Pan European Grid Advanced Simulation and State Estimation (PEGASE) project. Thanks to Cédric Josz and colleagues from the French Transmission System Operator.

New Documentation

• Added an online function reference to the website at http://www.pserc.cornell.edu/matpower/docs/ref/.

- Added support for using PARDISO (https://pardiso-project.org/) as linear solver for computing interior-point update steps in MIPS, resulting in dramatic improvements in computation time and memory use for very large-scale problems.
- Added support for LP/QP solver CLP (COIN_OR Linear Programming, https://github.com/coin-or/Clp). Use opf.dc.solver option 'CLP' or qps_clp().
- Added support for OPTI Toolbox (https://www.inverseproblem.co.nz/OPTI/) versions of CLP, GLPK and IPOPT solvers, providing a very simple installation path for some free high-performance solvers on Windows platforms.
- Network reduction toolbox for creating smaller approximate network equivalents from a larger original case file. *Contributed by Yujia Zhu and Daniel Tylavsky*.

⁸⁴https://matpower.org/docs/MATPOWER-manual-5.1.pdf

- Added unified interface to various solvers for mixed-integer linear and quadratic programming (MILP/MIQP) problems.
- Major update to have_fcn(), which now determines and caches version numbers and release dates for optional packages, and includes ability to toggle the availability of optional functionality.
- New and updated support for 3rd party solvers:
 - High-performance IPOPT-PARDISO solver builds from the PARDISO Project
 https://pardiso-project.org
 (at time of release this is MATPOWER's highest performing solver for very large scale AC OPF problems)
 - OPTI Toolbox versions of CLP, GLPK, IPOPT
 - CLP
 - Gurobi 6.x
 - Knitro 9.1
 - MOSEK 7.1
 - Optimization Toolbox 7.2
 - * dual-simplex algorithm for linprog()
 - * intlinprog() for MILP

• New functions:

- mplinsolve() provides unified interface for linear system solvers, including PARDISO and built-in backslash operator
- miqps_matpower() provides unified interface to multiple MILP/MIQP solvers.
- migps_clex() provides a unified MILP/MIQP interface to CPLEX.
- miqps_glpk() provides a unified MILP interface to GLPK.
- miqps_gurobi() provides a unified MILP/MIQP interface to Gurobi.
- migps_mosek() provides a unified MILP/MIQP interface to MOSEK.
- miqps_ot() provides a unified MILP interface to intlingprog().
- mosek_symbol() defines symbolic constants for setting MOSEK options.

Other Improvements

- Cleaned up and improved consistency of output in printpf() for generation and dispatchable load constraints.
- Modified runcpf() to gracefully handle the case when the base and target cases are identical (as opposed to getting lost in an infinite loop).
- Optional generator and dispatchable load sections in pretty-printed output now include off-line units.

Bugs Fixed

- Fixed fatal bug in case_info() for islands with no generation.
- Fixed fatal bug in toggle_dcline() when pretty-printing results. Thanks to Deep Kiran for reporting.
- Fixed sign error on multipliers on lower bound on constraints in qps_clp() and qps_glpk().
- Fixed bug in handling of interface flow limits, where multipliers on binding interface flow limits were off by a factor of the p.u. MVA base.
- Fixed minor bug with poly2pwl(), affecting units with PMAX ≤ 0 .
- Fixed error in qps_mosek() in printout of selected optimizer when using MOSEK 7.
- Fixed bug in hasPQcap() that resulted in ignoring generator capability curves for units whose reactive range increases as real power output increases. *Thanks to Irina Boiarchuk for reporting.*
- Fixed several incompatibilities with MATLAB versions < 7.3.

H.11 Version 6.0 – released Dec 16, 2016

The Matpower 6.0 User's Manual is available online.⁸⁵

New Open Development Model

- MATPOWER development has moved to GitHub! The code repository is now publicly available to clone and submit pull requests.⁸⁶
- Public issue tracker for reporting bugs, submitting patches, etc. 87
- Separate repositories for MATPOWER, MOST, MIPS, MP-Test, all available from https://github.com/MATPOWER/.
- New developer e-mail list (MATPOWER-DEV-L) to facilitate communication between those collaborating on MATPOWER-related development. Sign up at: https://matpower.org/mailing-lists/#devlist.

New Case Files

- Added 9 new case files, 8 cases ranging from 1888 to 6515 buses representing the French system, and a 13,659-bus case representing parts of the of the European high voltage transmission network, stemming from the Pan European Grid Advanced Simulation and State Estimation (PEGASE) project. Thanks again to Cédric Josz and colleagues from the French Transmission System Operator. Please cite reference [45] when publishing results based on these cases.
- Added case145.m, IEEE 145 bus, 50 generator dynamic test case from the U of WA Power Systems Test Case Archive⁸⁸.
- Added case33bw.m, a 33-bus radial distribution system from Baran and Wu.

⁸⁵https://matpower.org/docs/MATPOWER-manual-6.0.pdf

⁸⁶https://github.com/MATPOWER/matpower

⁸⁷https://github.com/MATPOWER/matpower/issues

⁸⁸https://labs.ece.uw.edu/pstca/dyn50/pg_tcadd50.htm

- Matpower Optimal Scheduling Tool (MOST) 1.0b1 is a major new feature, implementing a full range of optimal power scheduling problems, from a simple as a deterministic, single period economic dispatch problem with no transmission constraints to as complex as a stochastic, security-constrained, combined unit-commitment and multiperiod OPF problem with locational contingency and load-following reserves, ramping costs and constraints, deferrable demands, lossy storage resources and uncertain renewable generation. See docs/MOST-manual.pdf for details.
- General mechanism for applying modifications to an existing MATPOWER case. See apply_changes() and idx_ct().
- Redesigned CPF callback mechanism to handle CPF events such as generator limits, nose point detection, etc. Included event log in CPF results.
- Added options 'cpf.enforce_p_lims' and 'cpf.enforce_q_lims' to enforce generator active and reactive power limts in the continuation power flow.
- Added OPF option 'opf.use_vg' to provide a convenient way to have the OPF respect the generator voltage setpoints specified in the gen matrix.
- Experimental foundation for handling of ZIP load models in power flow (Newton, fast-decoupled only), continuation power flow, and optimal power flow (MIPS, fmincon, Knitro, IPOPT solvers only). Currently, ZIP loads can only be specified on a system-wide basis using the experimental options 'exp.sys_wide_zip_loads.pw' and 'exp.sys_wide_zip_loads.qw'.
- Support for quadprog() under GNU Octave.
- New contributed extras:
 - Plot electrically meaningful drawings of a MATPOWER case using plot_mpc()
 in extras/misc, contributed by Paul Cuffe.
 - Find the maximum loadability limit of a system via an optimal power flow and dispatchable loads, using maxloadlim() in extras/maxloadlim, contributed by Camille Hamon.
 - Create a quadratically-constrained quadratic programming (QCQP) representation of the AC power flow problem using using qcqp_opf() in extras/misc, contributed by Cédric Josz and colleagues.

• New functions:

- apply_changes() and idx_ct() provide a general mechanism for applying modifications to an existing MATPOWER case.
- feval_w_path() evaluates a function located at a specified path, outside of the MATLAB path.
- mpopt2qpopt() provides a common interface for creating options struct for mi/qps_matpower() from a MATPOWER options struct.

• New function options:

- Option to call makeB(), makeBdc(), makePTDF(), scale_load(), and total_load() with full case struct (mpc) instead of individual data matrices (bus, branch, etc.).
- total_load(), which now computes voltage-dependent load values, accepts the values 'bus' and 'area' as valid values for 'load_zone' argument.

Other Improvements

- Changed default solver order for LP, QP, MILP, MIQP problems to move Gurobi before CPLEX and BPMPD after Optimization Toolbox and GLPK.
- Added some caching to mpoption() and made minor changes to nested_struct_copy() to greatly decrease the overhead added by mpoption() when running many small problems.
- Added option 'cpf.adapt_step_damping' to control oscillations in adaptive step size control for continuation power flow.
- Added CPF user options for setting tolerances for target lambda detection and nose point detection, 'cpf.target_lam_tol' and 'cpf.nose_tol', respectively.
- Added support for Matlab Optimization Toolbox 7.5 (R2016b).
- Added support for MOSEK v8.x.
- Added tests for power flow with 'pf.enforce_q_lims' option.
- Updated network reduction code to handle cases with radially connected external buses.

- Updated versions of qcqp_opf() and qcqp_opf() in extras/misc, from Cédric Josz.
- Added "Release History" section to Appendix of manual.
- Many new tests.

Bugs Fixed

- Fixed bug in toggle_dclines() that resulted in fatal error when used with OPF with reactive power costs. *Thanks to Irina Boiarchuk*.
- Fixed fatal bug in update_mupq() affecting cases where QMIN is greater than or equal to QC1MIN and QC2MIN (or QMAX is less than or equal to QC1MAX and QC2MAX) for all generators. Thanks Jose Miquel.
- Copying a field containing a struct to a non-struct field with nested_struct_copy() now overwrites rather than causing a fatal error.
- Fixed a bug in psse_convert_xfmr() where conversion of data for transformers with CZ=3 was done incorrectly. Thanks to Jose Luis Marín and Yujia Zhu.
- Fixed a fatal bug in psse_convert_xfmr() affecting transformers with CW and/or CZ equal to 1. Thanks to Matthias Resch.
- Fixed a crash in have_fcn() caused by changes in OPTI Toolbox v2.15 (or possibly v2.12)
- Commented out isolated bus 10287 in case3375wp.m.
- Added code to DC OPF to return success = 0 for cases where the matrix is singular (e.g. islanded system without slack).
- Fixed problem in have_fcn() where SeDuMi was turning off and leaving off all warnings.
- Fixed shadow prices on variable bounds for AC OPF for fmincon, IPOPT, and Knitro.
- In savecase() single quotes are now escaped properly in bus names.
- Generator capability curve parameters that define a zero-reactive power line no longer cause a fatal error.

• Bad bus numbers no longer cause a fatal error (after reporting the bad bus numbers) in case_info().

Incompatible Changes

- Removed fairmax() from the public interface by moving it inside uopf(), the only place it was used.
- Removed 'cpf.user_callback_args' option and modified 'cpf.user_callback'.
- Changed name of 'cpf.error_tol' option to 'cpf.adapt_step_tol'.
- Changed the Sbus argument in newtonpf() from a vector to a function handle to a function that returns a vector.

H.12 Version 7.0 – released Jun 20, 2019

The MATPOWER 7.0 User's Manual is available online.⁸⁹

- New Matpower installer script install_matpower() automatically updates Matlab or Octave paths or optionally provides the commands required to so.
- Support for additional user-defined general nonlinear constraints and costs in AC OPF.
- Support for exporting Matpower case to PSS/E RAW data format.
- Three new variants of the standard Newton AC power flow, for a total of four, including both nodal power and current balance constraints and both polar and cartesian representations of voltage. See the new pf.current_balance and pf.v_cartesian options. Thanks to Baljinnyam Sereeter.
- Three new variants of the standard AC OPF formulation, for a total of four, including both nodal power and current balance constraints and both polar and cartesian representations of voltage. See the new opf.current_balance and opf.v_cartesian options. Thanks to Baljinnyam Sereeter.
- Three new power flow algorithms for radial distribution systems selected via the three new options for pf.alg, namely 'PQSUM', 'ISUM', 'YSUM'. Also includes new MATPOWER options pf.radial.max_it and pf.radial.vcorr. See Section 4.3 on "Distribution Power Flow" for details. Thanks to Mirko Todorovski.
- Major upgrade to OPF soft limit functionality, supporting soft limits on all AC and DC OPF inequality constraints, including branch flow constraints, bus voltage bounds, generator active and reactive bounds, branch flow and branch angle difference limits. Thanks to Eran Schweitzer.
- Docker image tagged matpower/matpower-desktop, providing a pre-packaged Ubuntu desktop environment with Octave, Matpower, and the Matpower Extras all pre-installed. See the docker/Matpower-Docker.md page for more details. Thanks to Richard Lincoln.

⁸⁹https://matpower.org/docs/MATPOWER-manual-7.0.pdf

• New options:

- pf.nr.lin_solver controls the linear solver used to compute the Newton update step in the Newton-Raphson power flow.
- pf.radial.max_it and pf.radial.vcorr are options for the new radial power flow algorithms. Thanks to Mirko Todorovski.
- cpf.enforce_flow_lims and cpf.enforce_v_lims control enforcement of branch flow and bus voltage magnitude limits in the continuation power flow and cpf.flow_lims_tol and cpf.v_lims_tol control the respective detection tolerances. Thanks to Ahmad Sadiq Abubakar and Shrirang Abhyankar.
- pf.current_balance and pf.v_cartesian control formulation used for Newton AC power flow. Thanks to Baljinnyam Sereeter.
- opf.current_balance and opf.v_cartesian control formulation used for AC OPF. Thanks to Baljinnyam Sereeter.
- opf.softlims.default determines whether or not to include soft limits on constraints whose parameters are not specified explicitly in the mpc.softlims struct. For use with enhanced toggle_softlims() functionality. Thanks to Eran Schweitzer.
- opf.start replaces deprecated opf.init_from_mpc and adds a new possibility to automatically run a power flow to initialize the starting state for the OPF.

• New functions:

- calc_branch_angle calcultes voltage angle differences for branches.
- dImis_dV() evaluates the partial derivatives of nodal current balance with respect to bus voltages.
- d2Imis_dV2() evaluates the 2nd derivatives of nodal current balance with respect to bus voltages.
- d2Imis_dVdSg() evaluates the 2nd derivatives of nodal current balance with respect to bus voltages and generator injections.
- d2Abr_dV2() evaluates the 2nd derivatives of squared branch flows with respect to bus voltages.
- gentypes() and genfuels provide list of standard generator unit types and fuel types, respectively.

- install_matpower() installs MATPOWER by automatically modifying or, alternatively, showing needed modifications to MATLAB or Octave path.
- loadshed() computes MW curtailments of dispatchable loads.
- opf_branch_ang_fcn() evaluates AC branch flow limit constraints and gradients.
- opf_branch_ang_hess() evaluates Hessian of AC branch flow limit constraints.
- opf_current_balance_fcn() evaluates AC current balance constraints and gradients.
- opf_current_balance_hess() evaluates Hessian of AC current balance constraints.
- opf_veq_fcn() evaluates voltage magnitude equality constraints and gradients.
- opf_veq_hess() evaluates Hessian of voltage magnitude equality constraints.
- opf_vlim_fcn() evaluates voltage magnitude limit constraints and gradients.
- opf_vlim_hess() evaluates Hessian of voltage magnitude limit constraints.
- opf_vref_fcn() evaluates reference voltage angle equality constraints and gradients.
- opf_vref_hess() evaluates Hessian of reference voltage angle equality constraints.
- opt_model/add_lin_constraint() to add linear constraints to an optimization model.
- opt_model/add_nln_constraint() to add nonlinear constraints to an optimization model.
- opt_model/init_indexed_name() to initialize the indices for an indexed name set of constraints, costs or variables.
- save2psse() to export a Matpower case to PSS/E RAW data format.
- savechgtab() to save change tables, such as those used by apply_changes,
 to a file.

New Case Files

- Seven new purely synthetic cases from the ACTIVSg team (ASU, Cornell, Texas A&M, U of Illinois, and VCU Synthetic grids), resulting from work supported by the ARPA-E GRID DATA program. Thanks to Adam Birchfield and the ACTIVSg team.
 - case_ACTIVSg200 200-bus Illinois synthetic model
 - case_ACTIVSg500 500-bus South Carolina synthetic model
 - case_ACTIVSg2000 2000-bus Texas synthetic model
 - case_ACTIVSg10k 10,000-bus US WECC synthetic model
 - case_ACTIVSg25k 25,000-bus US Northeast/Mid-Atlantic synthetic model
 - case_ACTIVSg70k 70,000-bus Eastern US synthetic model
 - case_SyntheticUSA 82,000-bus continental USA synthetic model (aggregation of case_ACTIVSg70k, case_ACTIVSg10k, and case_ACTIVSg2000, connected by 9 DC lines)

Some of these cases also include contingency tables and/or hourly load scenarios for 1 year.

- contab_ACTIVSg200
- contab_ACTIVSg500
- contab_ACTIVSg2000
- contab_ACTIVSg10k
- scenarios_ACTIVSg200
- scenarios_ACTIVSg2000
- New RTS-GMLC case from https://github.com/GridMod/RTS-GMLC.
 - case4_RTS_GMLC
- Six new radial distribution system cases. Thanks to Mirko Todorovski.
 - case4_dist
 - case18
 - case22
 - case69
 - case85
 - case141

New Documentation

- Two new Tech Notes, available from MATPOWER home page.
 - MATPOWER *Technical Note 3* "Addendum to AC Power Flows and their Derivatives using Complex Matrix Notation: Nodal Current Balance" [39]
 - Matpower Technical Note 4 "AC Power Flows and their Derivatives using Complex Matrix Notation and Cartesian Coordinate Voltages" [40]
- LaTeX source code for MATPOWER User's Manual included in docs/src, for MIPS User's Manual in mips/docs/src and for MOST User's Manual in most/docs/src.

New Website

• MATPOWER has a new website at https://matpower.org. Please discontinue use of the old http://www.pserc.cornell.edu/matpower/ address.

Other Improvements

- Update versions of included packages:
 - MIPS 1.3.1.
 - MOST 1.0.2.
 - MP-Test 7.0.
- Continuous integration testing via GitHub and Travis-CI integration.
- Support added in core optimization model opt_model for:
 - general nonlinear constraints
 - general nonlinear costs
 - quadratic costs
- Refactor OPF code to take advantage of new opt_model capabilities for nonlinear constraints and quadratic and nonlinear costs.
- Derivative functions now support cartesian coordinates for voltage in addition to polar coordinates.

- In the Newton power flow, for larger systems use explicit LU decomposition with AMD reordering and the 3 output argument form of lu (to select the Gilbert-Peierls algorithm), resulting in up to a 2x speedup in MATLAB, 1.1x in Octave. Thanks to Jose Luis Marín.
- Improve performance of makeYbus(). Thanks to Binbin Chen.
- Support plotting of multiple nose curves in CPF by allowing option cpf.plot.bus to take on vector values.
- Add line for curtailed load to case_info() output.
- Change default implementation of active power line flow constraints (opf.flow_lim = 'P') to use flow directly, rather than square of flow, which is now a separate option, namely opf.flow_lim = '2'. Thanks to Nico Meyer-Huebner.
- Add genfuels and gentypes to establish standard set of values for optional mpc.genfuel and mpc.gentype fields for generator fuel type and generator unit type, respectively.
- Add support for gentype and genfuel fields of MATPOWER case struct in extract_islands, ext2int, int2ext, load2disp and savecase.
- Add support for bus_name field of MATPOWER case struct to extract_islands, ext2int and int2ext.
- Add support for IPOPT solver under Octave, including in the Travis-CI testing. Thanks to Richard Lincoln.
- Add support for YALMIP, SeDuMi and SDPT3 to be recognized under Octave. Thanks to Richard Lincoln.
- Add 'clear_cache' options to have_fcn() (see issue #65) to facilitate rechecking for optional functionality after changes to the MATLAB/Octave path.
- Deprecated functions:
 - d2AIbr_dV2() use dA2br_dV2() instead.
 - d2ASbr_dV2() use dA2br_dV2() instead.
 - opt_model/add_constraints() use the corresponding one of the following methods instead: add_lin_constraint(), add_nln_constraint(), or init_indexed_name().

- opt_model/add_costs() use the corresponding one of the following methods instead: add_quad_cost(), add_nln_cost(), add_legacy_cost(), or init_indexed_name().
- opt_model/linear_constraints() use opt_model/params_lin_constraint()
 instead.
- opt_model/build_cost_params() no longer needed, incorporated into opt_model/params_legacy_cost().
- opt_model/get_cost_params() use opt_model/params_legacy_cost() instead.

Bugs Fixed

- Fix bug in conversion of older versions of Matpower options.
- Fix bug #4 where some Q limits were not being respected by CPF when buses were converted to PQ by initial power flow run. Thanks to Shruti Rao.
- Fix fatal bug #8 when calling runcpf with base and target cases with identical load and generation. Thanks to Felix.
- Fix fatal bug in get_losses when computing derivatives of reactive branch injections and fix some related tests.
- Fix #11 fatal error encountered when running test_matpower with SDP_PF and YALMIP installed, but no SDP solver. Now checks for availability of SeDuMi, SDP3 or MOSEK before attempting to run SDP_PF tests that require solving an SDP. Thanks to Felix.
- Fix bug #12 where the CPF could terminate early when requesting trace of the full curve with P or Q limits enforced, if a limit becomes binding at the base case. Thanks to Felix.
- Fix bug #13 where setting all buses to type NONE (isolated) resulted in a fatal error for ext2int, runpf, runcpf and runopf. Thanks to SNPerkin.
- Fix bug #21 where a continuation power flow that failed the first corrector step would produce a fatal error. Thanks to Elis Nycander.
- Fix bug #23 where the continuation power flow could switch directions unexpectedly when the operating point switched from stable to unstable manifold

or vice-versa after hitting a limit. Thanks to Elis Nycander and Shrirang Abhyankar.

- Fix bug #26 where, in a continuation power flow, a reactive limit at a bus could be detected in error if multiple generators at the bus had reactive ranges of very different sizes. Thanks to Elis Nycander and Shrirang Abhyankar.
- Fix runpf handling of case where individual power flow fails during Q limit enforcement.
- Fix bug #53 where certain OPF cases (e.g. case33bw) resulted in a fatal error under versions of MATLAB prior to R2016b (v9.1). Thanks to Jane Cheung.
- Fix bug #56 where save2psse was missing entries for two transformer columns, namely, VMA1 and VMI1. Thanks to Amin Gholami.
- Fix bug #57 where save2psse always used 1 for the CKT number, even for parallel branches or transformers. Thanks to Amin Gholami.
- Fix bug in have_fcn() where it would incorrectly mark Gurobi as available even if it had an expired license or failed for some other reason.
- Fix issue #60 by adding missing generator at slack bus in RTE cases. Affects the following cases:
 - case1888rte
 - case1951rte
 - case2848rte
 - case2868rte
 - case6468rte
 - case6470rte
 - case6495rte
 - case6515rte

Thanks to Jean Maeght.

Incompatible Changes

- Move included MATPOWER case files to new data subdirectory.
- Turning soft limits on without specifying any parameters explicitly in mpc.softlims now implements soft limits for all constraints, by default, not just branch flow limits. And the format of the input parameters in mpc.softlims has changed. See help toggle_softlims or Tables 7-9, 7-10 and 7-11 for the details.
- Swap the order of the output arguments of dSbus_dV()) for polar coordinate voltages (angle before magnitude) for consistency.
- Correct signs of phase shifter angles in Polish system cases, since they were based on the old sign convention used by MATPOWER prior to v3.2 (see change on 6/21/07). Affects the following cases:
 - case2383wp
 - case2736sp
 - case2737sop
 - case2746wop
 - case2746wp
 - case3375wp

Thanks to Mikhail Khokhlov and Dr. Artjoms Obusevs for reporting.

- Remove nln.mu.l.<name> and nln.mu.u.<name> fields from OPF results struct.

 Use nle.lambda.<name> and nli.mu.<name> fields instead for nonlinear constraint multipliers.
- Modify order of default output arguments of opt_model/get_idx().
- Add mpopt to input args for OPF 'ext2int', 'formulation', and 'int2ext' callbacks.
- Eliminate unnecessary reordering of on-line generators (sorted by increasing bus index) from ext2int(). The order is now left unmodified by ext2int(). This change should only affect user code that explicitly depends on the order of generators with internal numbering (hopefully quite rare).

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