# $ExaGO^{TM}$ Users Manual

Version 1.0

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# Chapter 1

# Introduction

The Exascale Grid Optimization (ExaGO<sup>TM</sup>) toolkit is an open source package for solving large-scale power grid optimization problems on parallel and distributed architectures, particularly targeted for exascale machines with heteregenous architectures (GPU). ExaGO<sup>TM</sup> is written in C/C++ and makes heavy use of the functionality provided by the PETSc[1] library. It uses RAJA [3] and Umpire [4] libraries for execution on the GPU and can make use of several optimization solvers - Ipopt [7], HiOp [6, 5], and TAO [1]- for solving the optimization problems.

All  $\rm ExaGO^{\rm TM}$  applications use a nonlinear formulation based on full AC optimal power flow. The different applications available with ExaGO are listed in Table 1.1

Table 1.1: ExaGO applications

Application	Description	Notes
OPFLOW	AC optimal power flow	
SCOPFLOW	Security-constrained AC optimal power	Uses TCOPFLOW for multi-
	flow	period contingencies
TCOPFLOW	Multi-period AC optimal power flow	
SOPFLOW	Stochastic AC optimal power flow	Uses SCOPFLOW for multi-
		contingency scenarios

While ExaGO is targeted for making use of distributed computing environments and GPUs, its full support is still under development. Table 1.2 lists the architecture support for ExaGO applications.

Table 1.2: ExaGO application execution on different hardware

Application	CPU (serial)	CPU (parallel)	GPU
OPFLOW	Y	N	Y
SCOPFLOW	Y	Y (embarrasingly parallel)	N
SOPFLOW	Y	Y (embarrasingly parallel)	N
TCOPFLOW	Y	N	N

Together, these applications cater to problems spanning the dimensions of security (contingencies), stochasticity, and time as shown in Figure 1.1.

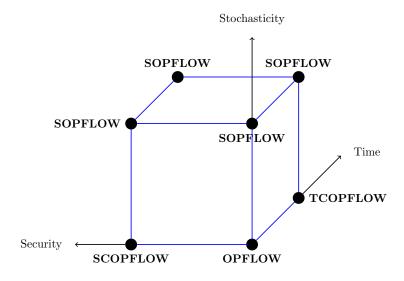


Figure 1.1: ExaGO provides applications along the dimensions of security (contingencies), time, and stochasticity. The label vertices denote different ExaGO applications available.

# Chapter 2

# Getting Started

#### 2.1 System requirements

ExaGO is currently only built on 64b OSX and Linux machines, compiled with GCC >= 7.3. We build ExaGO on Intel and IBM Power9 architectures.

#### 2.2 Prerequisites

This section assumes that you already have the ExaGO source code, and that the environment variable EXAGODIR is the directory of the ExaGO source code. ExaGO may be acquired via the PNNL git repository linked here, like so:

```
> git clone https://gitlab.pnnl.gov/exasgd/frameworks/exago.git exago > export EXAGODIR=$PWD/exago
```

Paths to installations of third party software in examples are abbreviated with placeholder paths. For example, /path/to/cuda is a placeholder for a path to a valid CUDA Toolkit installation.

# 2.3 Dependencies

ExaGO has dependencies in table 2.1.

Table 2.1: Dependency Table

Dependency	Version Constraints	Mandatory	Notes
PETSc [1]	>= 3.13.0	✓	Only needed for the setup stage
CMake	>= 3.10	✓	Only a build dependency
MPI	>= 3.1.3		Only tested with openmpi and spectrummpi
Ipopt [7]	>=3.12		
HiOp [6, 5]	>=0.3.0		Prefer dynamically linked
RAJA [3]	>= 0.11.0		
Umpire [2]	>= 2.1.0		Only when RAJA is enabled
MAGMA	>= 2.5.2		Only when GPU acceleration is enabled
CUDA Toolkit	>= 10.2.89		Only when GPU acceleration is enabled

These may all be toggled via CMake, which will be discussed in the section Building and installation.

#### 2.3.1 Notes on environment modules

Many of the dependencies are available via environment modules on institutional clusters. To get additional information on your institution's clusters, please ask your institution's system administrators. Some end-to-end examples in this document will use system-specific modules and are not expected to expected to run on other clusters.

For example, the modules needed to build and run ExaGO on Newell, an IBM Power9 PNNL cluster, are as follows:

```
> module load gcc/7.4.0
> module load openmpi/3.1.5
> module load cuda/10.2
> module load magma/2.5.2_cuda10.2
> module load metis/5.1.0
> module load cmake/3.16.4
```

#### 2.3.2 Additional Notes on GPU Accelerators

As of January 2021, CUDA is the only GPU accelerator platform ExaGO fully supports. We have preliminary support for HIP, however this is not in any of our main development branches yet. Full HIP support should arrive in early 2021.

#### 2.3.3 Additional Notes on Umpire

Umpire is an implicit dependency of RAJA. If a user enables RAJA, they must also provide a valid installation of Umpire. Additionally, if a user would like to run ExaGO with RAJA and without CUDA, they must provide a CPU-only build of Umpire since an Umpire build with CUDA enabled will link against CUDA.

## 2.4 Building and installation

#### 2.4.1 Default Build

ExaGO may be built with a standard CMake workflow:

Listing 2.1: Example CMake workflow

```
> cd $EXAGODIR
> export BUILDDIR=$PWD/build INSTALLDIR=$PWD/install
> mkdir $BUILDDIR $INSTALLDIR
> cd $BUILDDIR
> cmake .. -DCMAKE_INSTALL_PREFIX=$INSTALLDIR
> make install
```

The following sections will assume the user is following the basic workflow outlined above.

Note: For changes to the CMake configuration to take effect, the code will have to be rebuilt.

#### 2.4.2 Additional Options

To enable additional options, CMake variables may be defined via CMake command line arguments, ccmake, or cmake-gui. CMake options specific to ExaGO have an EXAGO\_ prefix. For example, the following shell commands will build ExaGO with MPI:

```
> cmake .. -DCMAKE_INSTALL_PREFIX=$INSTALLDIR -DEXAGO_ENABLE_MPI=ON
```

ExaGO's CMake configuration will search the usual system locations for an MPI installation.

For dependencies not installed to a system-wide location, users may also directly specify the location of a dependency. For example, this will build ExaGO with IPOPT enabled and installed to a user directory:

```
> cmake .. \
    -DCMAKE_INSTALL_PREFIX=$INSTALLDIR \
    -DEXAGO_ENABLE_IPOPT=ON \
    -DIPOPT_DIR=/path/to/ipopt
```

Notice that the CMake variable IPOPT\_DIR does not have an EXAGO\_ prefix. This is because the variables specifying locations often belong to external CMake modules. CMake variables indicating installation directories do not have an EXAGO\_ prefix.

Some CMake options effect others. This is especially common when the user enables ExaGO's GPU options. For example, if the user enables EXAGO\_ENABLE\_GPU and EXAGO\_ENABLE\_RAJA, the user must provide a GPU-enabled RAJA installation. Umpire is also an implicit dependency of RAJA, so if the user enables EXAGO\_ENABLE\_GPU they must also provide a GPU-enabled Umpire installation.

Below is a complete shell session on PNNL's cluster Newell in which a more complicated ExaGO configuration is built, where each dependency installation is explicitly passed to CMake. Environment modules specific to Newell are provided to make the example thorough, even though they are not likely to work on another machine.

Listing 2.2: ExaGO build with all options enabled

```
> module load gcc/7.4.0
> module load cmake/3.16.4
> module load openmpi/3.1.5
> module load magma/2.5.2_cuda10.2
> module load metis/5.1.0
> module load cuda/10.2
> git clone https://gitlab.pnnl.gov/exasgd/frameworks/exago.git exago
> export EXAGODIR=$PWD/exago
> cd $EXAGODIR
> export BUILDDIR=$PWD/build INSTALLDIR=$PWD/install
> mkdir $BUILDDIR $INSTALLDIR
> cd $BUILDDIR
> cmake .. \
  -DCMAKE_INSTALL_PREFIX=$INSTALLDIR \
  -DCMAKE BUILD TYPE=Debug \
  -DEXAGO_ENABLE_GPU=ON \
  -DEXAGO_ENABLE_HIOP=ON \
  -DEXAGO_ENABLE_IPOPT=ON \
```

```
-DEXAGO_ENABLE_MPI=ON \
  -DEXAGO_ENABLE_PETSC=ON \
  -DEXAGO_RUN_TESTS=ON \
  -DEXAGO_ENABLE_RAJA=ON \
  -DEXAGO_ENABLE_IPOPT=ON \
  -DIPOPT_DIR=/path/to/ipopt \
  -DRAJA_DIR=/path/to/raja \
  -Dumpire_DIR=/path/to/umpire \
  -DHIOP_DIR=/path/to/hiop \
  -DMAGMA_DIR=/path/to/magma \
  -DPETSC DIR=/path/to/petsc
> make -j 8 install
 # For the following commands, a job scheduler command may be needed.
> # Run test suite
> make test
> # Run an ExaGO application:
> $INSTALLDIR/bin/opflow
```

#### 2.5 Usage

Each ExaGO application has the following format for execution

```
./app <app_options>
```

Here, app\_options are the command line options for the application. Each application has many options through which the input files and the control options can be set for the application. All application options have the form -app\_option\_name followed by the app\_option\_value. For instance,

```
./opflow -netfile case9mod.m -opflow_model POWER_BALANCE_POLAR \
-opflow_solver IPOPT
```

will execute the OPFLOW application using case9mod.m input file with the model POWER\_BALANCE\_POLAR

→ and Ipopt [7]solver.

Options can also be passed to each application through -options\_file, or through a combination of the command line and options file. The configuration specified last on the command line overrides any previous options. For example, if -options\_file opflowoptions specified -netfile case9mod.m within it's settings:

```
./opflow -netfile case118.m -options_file opflowoptions # Uses case9mod.m ./opflow -options_file opflowoptoins -netfile case118.m # Uses case118.m
```

If no options file is specified through the command line, ExaGO applications will attempt to locate the default options file for a given application in ./, ./options, <install\_dir>/options. Each solver that can be used within ExaGO also have their own unique options. Some of these options are specified internally within ExaGO, but any other options that are needed should either be passed through the command line, or additionally specified in the options file.

# Chapter 3

# Optimal power flow (OPFLOW)

OPFLOW solves the full AC optimal power flow problem and provides various flexible features that can be toggled via run-time options. It has interfaces to different optimization solvers and includes different representations of the underlying equations (power-balance-polar and power-balance-cartesian) that can be used. By selecting the appropriate solver, OPFLOW can be executed on CPUs (in serial or parallel) or on GPUs.

#### 3.1 Formulation

Optimal power flow is a general nonlinear programming problem with the following form

$$\min f(x) \tag{3.1}$$

s.t.

$$g(x) = 0 (3.2)$$

$$h(x) \le 0 \tag{3.3}$$

$$x^{\min} \le x \le x^{\max} \tag{3.4}$$

Here, x are the decision variables with lower and upper bounds  $x^{\min}$  and  $x^{\max}$ , respectively, f(x) is the objective function, g(x) and h(x) are the equality and inequality constraints, respectively. In the following sections we describe what constitutes these different terms as used by OPFLOW.

#### 3.1.1 Variables and bounds

The different variables used in OPFLOW formulation are described in Table 3.1.

Power imbalance variables are non-physical (slack) variables that measure the violation of power balance at buses. Having these variables (may) help in making the optimization problem easier to solve since they always ensure feasibility of the bus power balance constraints.

#### 3.1.2 Objective Function

The objective function for OPFLOW is given in (3.5)

min. 
$$C(p^{g}) + C(\Delta p^{g}) + C(\Delta p^{l}, \Delta q^{l}) + C(\Delta p, \Delta q)$$
 (3.5)

Table 3.1: Optimal power flow (OPFLOW) variables

Variable	Symbol	Bounds	Notes
Generator real power dispatch	$p_j^{\mathrm{g}}$	$p_j^{\text{gmin}} \le p_j^{\text{g}} \le p_j^{\text{gmax}}$	
Generator reactive power dispatch	$q_j^{\mathrm{g}}$	$q_j^{\text{gmin}} \le q_j^{\text{g}} \le q_j^{\text{gmax}}$	
Generator real power deviation	$\Delta p_j^{ m g}$	$-p_j^{\mathrm{r}} \le \Delta p_j^{\mathrm{g}} \le p_j^{\mathrm{r}}$	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$
Generator real power set-point	$p_j^{\text{gset}}$	$\begin{array}{ccc} p_j^{\text{gmin}} & \leq & p_j^{\text{gset}} & \leq \\ p_j^{\text{gmax}} & & & \end{array}$	Only used when -opflow_has_gensetpoint or -opflow_use_agc option is active.
System power ex- cess/deficit	$\Delta P$	Unbounded	Only used when -opflow_use_agc is active
Bus voltage angle	$\theta_i$	$-\pi \le \theta_i \le \pi$	• Used with power balance polar model (-opflow_model POWER_BALANCE_POLAR) • $\theta_i$ is unbounded, except reference bus angle $\theta_i^{\text{ref}}$ which is fixed to 0
Bus voltage magnitude	$v_i$	$v_i^{\min} \le v_i \le v_i^{\max}$	• Used with power balance polar model (-opflow_model POWER_BALANCE_POLAR)  • $v_i^{\min} = v_i^{\max} = v_i^{\text{set}}$ if fixed generator set point voltage option is active (-opflow_has_gensetpoint)
Bus volt- age real part	$e_i$	$-v_i^{\max} \le e_i \le v_i^{\max}$	Used with power balance cartesian model (-opflow_model POWER_BALANCE_CARTESIAN)
Bus volt- age imagi- nary part	$f_i$	$-v_i^{\max} \le f_i \le v_i^{\max}$	Used with power balance cartesian model (-opflow_model POWER_BALANCE_CARTESIAN)
Bus real power mismatch	$\Delta p_i$	Unbounded	Used when power mismatch variable option is active (-opflow_include_powerimbalance_variables1)
Bus reactive power mismatch	$\Delta q_i$	Unbounded	Used when power mismatch variable option is active (-opflow_include_powerimbalance_variables1)
Real power load loss	$\Delta p_j^{\mathrm{l}}$	$0 \le \Delta p_j^{l} \le p_j^{l}$	Used when load loss variable option is active (-opflow_include_loadloss_variables1)
Reactive power load loss	$\Delta q_j^1$	$0 \le \Delta q_j^{\mathbf{l}} \le q_j^{\mathbf{l}}$	Used when load loss variable option is active (-opflow_include_loadloss_variables1)

#### Total generation cost $C(p^{\mathbf{g}})$

Needs option -opflow\_objective MIN\_GEN\_COST

$$C(p^{g}) = \sum_{j \in J^{\text{gen}}} C_{j}^{g}(p_{j}^{g})$$

$$(3.6)$$

Here,  $C_j^{\rm g}$  is a quadratic function of the form  $C_j^{\rm g} = a_j^{\rm g} p_j^{\rm g^2} + b_j^{\rm g} p_j^{\rm g} + c_j^{\rm g}$ .

#### Total generation setpoint deviation $C(\Delta p^{\mathbf{g}})$

Needs option -opflow\_objective MIN\_GENSETPOINT\_DEVIATION option

$$C(\Delta p^{g}) = \sum_{j \in J^{\text{gen}}} (\Delta p_{j}^{g^{2}})$$
(3.7)

#### Load loss $C(\Delta p^{\mathbf{l}}, \Delta q^{\mathbf{l}})$

This term gets added to the objective when -opflow\_include\_loadloss\_variables option is active.

$$C(\Delta p^{\mathrm{l}}, \Delta q^{\mathrm{l}}) = \sum_{j \in J^{\mathrm{ld}}} \sigma_{j}^{\mathrm{l}} (\Delta p_{j}^{\mathrm{l}^{2}} + \Delta q_{j}^{\mathrm{l}^{2}})$$

$$(3.8)$$

The load loss penalty  $\sigma_j^l$  can be set via the option -opflow\_loadloss\_penalty. The default is 1000/MW for all loads.

#### Power imbalance $C(\Delta p, \Delta q)$

This term gets added to the objective when -opflow\_include\_powerimbalance\_variables option is active.

$$C(\Delta p, \Delta q) = \sum_{i \in J^{bus}} \sigma_i (\Delta p_i^2 + \Delta q_i^2)$$
(3.9)

The power imbalance cost  $\sigma_i$  can be set via the option -opflow\_powerimbalance\_penalty. The default is \$10,000/MW for all buses. Though the power imbalance variables  $\Delta p_i, \Delta q_i$  are slack or non-physical, they can help in solving infeasible cases and provide a measure of the infeasibility.

#### 3.1.3 Equality constraints

#### Nodal power balance

$$\sum_{\substack{j \in J^{\text{gen}} \\ A^{\text{g}}_{ij} \neq 0}} p^{\text{g}}_{j} = p^{\text{sh}}_{i} + \sum_{\substack{j \in J^{\text{ld}} \\ A^{\text{l}}_{ij} \neq 0}} p^{\text{l}}_{j} + \sum_{\substack{j \in J^{\text{br}} \\ A^{\text{br}}_{oi} \neq 0}} p^{\text{br}}_{j_{od}} + \sum_{\substack{j \in J^{\text{br}} \\ A^{\text{br}}_{id} \neq 0}} p^{\text{br}}_{j_{do}}$$
(3.10)

$$\sum_{\substack{j \in J^{\text{gen}} \\ A_{ij}^{g} \neq 0}} q_{j}^{g} = q_{i}^{\text{sh}} + \sum_{\substack{j \in J^{\text{ld}} \\ A_{ij}^{l} \neq 0}} q_{j}^{l} + \sum_{\substack{j \in J^{\text{br}} \\ A_{oi}^{p} \neq 0}} q_{jod}^{\text{br}} + \sum_{\substack{j \in J^{\text{br}} \\ A_{oi}^{p} \neq 0}} q_{jdo}^{\text{br}}$$
(3.11)

(3.12)

where, the real and reactive power shunt consumption is given by (3.19) and (3.20)

The real and reactive power flow  $p_{j_{od}}^{\text{br}}$ ,  $q_{j_{od}}^{\text{br}}$  for line j from the origin bus o to destination bus d is given by (3.21) - (3.22) and from destination bus d to origin bus o is given by (3.23) - (3.24)

#### Generator real power output

When using -opflow\_has\_gensetpoint, two extra variables  $p_j^{\mathrm{gset}}$  and  $\Delta p_j^{\mathrm{g}}$  are added for each generator. The generator real power output  $p_i^{\rm g}$  is related to the power deviation  $\Delta p_i^{\rm g}$  by the following relations

$$p_j^{\text{gset}} + \Delta p_j^{\text{g}} - p_j^{\text{g}} = 0 \tag{3.13}$$

$$p_j^{\text{gset}} - p_j^{\text{g*}} = 0 (3.14)$$

The second equation here allows to fix the generator set point  $p_j^{\text{gset}}$  to a fixed value  $p_j^{\text{g}^*}$ . Here,  $p_j^{\text{g}^*}$  is the set-point for the generator real power output, which can be thought of as an operator set or contractual agreement set-point.

#### Inequality constraints

#### MVA flow on branches

MVA flow limits at origin and destination buses for each line.

$$p_{j_{od}}^{\text{br}^2} + q_{j_{od}}^{\text{br}^2} \le s_j^{\text{rateA}^2}, \quad j \in J^{\text{br}}$$

$$p_{j_{do}}^{\text{br}^2} + q_{j_{do}}^{\text{br}^2} \le s_j^{\text{rateA}^2}, \quad j \in J^{\text{br}}$$
(3.15)

$$p_{j_{do}}^{\text{br}^2} + q_{j_{do}}^{\text{br}^2} \le s_j^{\text{rateA}^2}, \quad j \in J^{\text{br}}$$
 (3.16)

To reduce the number of inequality constraints, only lines that are in service and having MVA A rating  $s_i^{\text{rateA}}$  less than 10000 MVA are considered.

#### Automatic generation control (AGC)

With -opflow\_use\_agc, two additional constraints are added for each participating generator to enforce the proportional generator redispatch participation as done in automatic generation control (AGC). These two equations are

$$\left(\alpha_j^{\mathrm{g}} \Delta P - \Delta p_j^{\mathrm{g}}\right) \left(p_j^{\mathrm{g}} - p_j^{\mathrm{gmax}}\right) \ge 0$$

$$\left(\left(\Delta p_j^{\mathrm{g}} - \alpha_j \Delta P\right) \left(p_j^{\mathrm{gmin}} - p_j^{\mathrm{g}}\right) \ge 0$$
(3.17)

Eq. 3.17 enforces the generator set-point deviation to be equal to the generation participation when the generator has head-room available  $p_j^{\text{gmin}} \leq p_j^{\text{g}} \leq p_j^{\text{gmax}}$ . Here,  $\alpha_j^{\text{g}}$  is the generator participation factor which is the proportion of the power deficit/excess  $\Delta P$  that the generator provides.

#### Generator bus voltage control

When the option -opflow\_qenbusvoltage FIXED\_WITHIN\_QBOUNDS is used, the generator bus voltage is fixed when the total reactive power generation available at the bus is within bounds. When it reaches its bounds, the voltage varies with the generator reactive power fixed at its bound. To implement this behavior, two inequality constraints are added for each generator bus

$$(v_i^{\text{set}} - v_i)(q_i - q^{\max_i}) \ge 0$$
  
 $(v_i - v_i^{\text{set}})(q^{\min_i} - q_i) \ge 0$  (3.18)

Here,  $q_i$ ,  $q^{\max_i}$ , and  $q^{\min_i}$  are the generated, maximum, and minimum reactive power at the bus, respectively.

#### 3.2Models

Table 3.2: OPFLOW models

Model type	OPFLOW option (-opflow_model )	Compatible solvers	CPU-GPU
Power balance with	POWER_BALANCE_POLAR	IPOPT	CPU
polar coordinates			
Power balance with	POWER_BALANCE_CARTESIAN	IPOPT, TAO	CPU
cartesian coordinates			
Power balance with	POWER_BALANCE_HIOP	HiOp [6, 5]	CPU/GPU
polar coordinates used			
with HIOP			
Power balance with	PBPOLRAJAHIOP	HiOp [6, 5]	GPU
polar coordinates used			
with HIOP on GPU			

#### 3.2.1Power balance polar

#### Shunt power

$$p_i^{\rm Sh} = g_i^{\rm Sh} v_i^2 \tag{3.19}$$

$$q_i^{\mathrm{Sh}} = -b_i^{\mathrm{Sh}} v_i^2 \tag{3.20}$$

#### **Branch flows**

The real and reactive power flow  $p_{j_{od}}^{\text{br}}$ ,  $q_{j_{od}}^{\text{br}}$  from bus o to bus d on line j is given by (3.21) – (3.22)

$$p_{j_{od}}^{\text{br}} = g_{oo}v_o^2 + v_o v_d (g_{od}\cos(\theta_o - \theta_d) + b_{od}\sin(\theta_o - \theta_d))$$

$$q_{j_{od}}^{\text{br}} = -b_{oo}v_o^2 + v_o v_d (-b_{od}\cos(\theta_o - \theta_d) + g_{od}\sin(\theta_o - \theta_d))$$
(3.21)

$$q_{j_{od}}^{\text{br}} = -b_{oo}v_o^2 + v_o v_d (-b_{od}\cos(\theta_o - \theta_d) + g_{od}\sin(\theta_o - \theta_d))$$
(3.22)

and from bus d to bus o is given by (3.23) - (3.24)

$$p_{j_{do}}^{\text{br}} = g_{dd}v_d^2 + v_d v_o (g_{do}\cos(\theta_d - \theta_o) + b_{do}\sin(\theta_d - \theta_o))$$
(3.23)

$$q_{j_{do}}^{\text{br}} = -b_{dd}v_d^2 + v_d v_o (-b_{do}\cos(\theta_d - \theta_o) + g_{do}\sin(\theta_d - \theta_o))$$
 (3.24)

Table 3.3: OPFLOW Model-solver compatibility

Model	Ipopt [7]	HiOp [6, 5]	TAO [1]
POWER_BALANCE_POLAR	✓		
POWER_BALANCE_CARTESIAN	✓		<b>√</b>
POWER_BALANCE_HIOP		✓	
PBPOLRAJAHIOP		✓	

#### 3.2.2 Power balance with HiOp

Power balance formulation with polar coordinates used with HiOp [6, 5][6] library. This model runs on CPU only, though the HiOp [6, 5]solver can run on GPU.

#### 3.2.3 Power balance with HiOp,RAJA,and Umpire

Power balance formulation with polar coordinates used with HiOp [6, 5] and RAJA [3]. This model uses RAJA [3] [3] and Umpire [4] libraries to run OPFLOW calculations (objective, constraints, etc.) on the GPU.

#### 3.3 Solvers

OPFLOW can be used with a few different solvers. All the three solvers solve the optimization problem via a nonlinear interior-point algorithm.

- 1. Ipopt [7] is a popular open-source package for solving nonlinear optimization problems. It is the most robust of the solvers implemented for solving OPFLOW. However, it can be run only on a single process and does not have GPU support. Option:
  - -opflow\_solver IPOPT
- 2. The Toolkit for Applied Optimization (TAO [1]) is a toolkit available through PETSc [1] for solving optimization problems. TAO includes a interior-point algorithm that is used by OPFLOW. The TAO algorithm is fully parallel, but is less robust compared to IPOPT. Currently, the TAO solver can only be used with the power balance cartesian model. Option:
  - -opflow\_solver TAO -opflow\_model POWER\_BALANCE\_CARTESIAN
- 3. HiOp [6, 5] is a high-performance optimization library that implements an interior-point algorithm for solving nonlinear optimization problems. The library supports execution both on the CPU and the GPU. Options:

```
CPU: -opflow_solver HIOP -opflow_model POWER_BALANCE_HIOP
```

GPU: -opflow\_solver HIOP -opflow\_model PBPOLRAJAHIOP

The option -hiop\_ipopt\_debug can be used with the CPU option in order to use an adapter that verifies the correctness of the first and second order derivates when compared to Ipopt [7]. This may be useful when debugging experimental changes.

## 3.4 Input and Output

The current ExaGO version only supports reading network file in MATPOWER format and can (optionally) write the output back in MATPOWER data file format.

# 3.5 Usage

./opflow -netfile <netfilename> <opflowoptions>

# 3.6 Options

See table 3.4

Table 3.4: OPFLOW options

Option	Meaning	Values (Default value)
-netfile	Network file name	string < 4096 characters
		(case9mod.m)
-print_output	Print output to screen	0 or 1 (0)
-save_output	Save output to file	0 or 1 (0)
-opflow_model	Representation of network bal-	See Table 3.2
	ance equations and bus voltages	(POWER_BALANCE_POLAR)
-opflow_solver	Optimization solver	See section 3.3
-opflow_initialization	Type of initialization	See Table 3.5 (MIDPOINT)
-opflow_has_gensetpoint	Uses generation set point and ac-	0 or 1 (0)
	tivates ramping variables	
-opflow_use_agc	Uses AGC formulation in OPF	0 or 1 (0)
-opflow_objective	type of objective function	See table 3.7 (MIN_GEN_COST)
-opflow_genbusvoltage	Type of generator bus voltage	See Table 3.6
	control	(FIXED_WITHIN_QBOUNDS)
-opflow_ignore_lineflow_constraints	Ignore line flow constraints	0 or 1 (0)
-opflow_include_loadloss_variables	Include load loss	0 or 1 (0)
-opflow_include_powerimbalance_variables	Include power imbalance	0 or 1 (0)
-opflow_loadloss_penality	\$ penalty for load loss	real (1000)
-opflow_powerimbalance_penalty	\$ penalty for power imbalance	real (10000)
-opflow_tolerance	Optimization solver tolerance	real (1e-6)

Table 3.5: OPFLOW initializations

Initialization type	Meaning
MIDPOINT	Use mid-point of bounds
FROMFILE	Use values from network file
ACPF	Run AC power flow for initialization
FLATSTART	Flat-start

Table 3.6: OPFLOW generator bus voltage control modes

Voltage control type	Meaning
FIXED_AT_SETPOINT	Fixed at given set-point. Reactive power limits are ignored
FIXED_WITHIN_QBOUNDS	Fixed within reactive power bounds
VARIABLE_WITHIN_BOUNDS	Variable within voltage bounds

Table 3.7: OPFLOW objective function types

Objective function	Meaning
MIN_GEN_COST	Minimize generation cost
MIN_GENSETPOINT_DEVIATION	Minimize deviation (ramp up-down) from generataor set-point
NO_OBJ	No objective function

### 3.7 Examples

Some OPFLOW example runs are provided with some sample output. Options values are the default values in table 3.4 unless otherwise specified. -print\_output is only used in the first example to save space. Sample output is generated by running examples from the installation directory.

Example using the TAO [1] solver:

```
$ ./bin/opflow -netfile case9mod.m -opflow_solver TAO -opflow_model \
POWER_BALANCE_CARTESIAN -print_output
[ExaGO INFO]: -- Checking ... -options_file not passed
                                                   exists: no
[ExaGO INFO]: Creating OPFlow
TAO solve converged due to CONVERGED_GRTOL iterations 12
______
              Optimal Power Flow
                 _____
Model
                                POWER_BALANCE_CARTESIAN
Solver
                                TAO
Objective
                                MIN GEN COST
Initialization
                                MIDPOINT
Load loss allowed
                                NO
Power imbalance allowed
                                NO
Ignore line flow constraints
                                NO
Number of variables
                                24
Number of equality constraints
                                19
Number of inequality constraints
                                27
Convergence status
                                CONVERGED
Objective value
                                4144.46
```

-----

Bus	Pd	Qd	Vm	Va	mult_Pmis	mult_Qmis
1	0.00	0.00	1.100	-0.000	2102.91	0.00
2	0.00	0.00	1.095	3.927	2059.18	0.00
3	0.00	0.00	1.087	2.120	2065.15	-0.00
4	0.00	0.00	1.097	-1.993	2103.16	0.08
5	75.00	50.00	1.079	-3.060	2113.46	7.29
6	90.00	30.00	1.087	-3.927	2129.85	1.62
7	0.00	0.00	1.100	0.535	2059.57	-0.04
8	100.00	35.00	1.089	-1.720	2079.34	2.99

9	0.00	0.00 1.1	00 -0.13	55 20	65.43	-0.09	
From	То			Stf		mult_Sf	mult_St
1	4	1	73.17	72.98	380.00	0.00	0.00
2	7	1	114.18	114.68	250.00	0.00	0.00
3	9	1	83.57	84.60	300.00	0.00	0.00
4	5	1	29.68	40.50	250.00	0.00	0.00
4	6	1	44.86	46.03	250.00	0.00	0.00
5	7	1	51.29	49.04	250.00	0.00	0.00
6	9	1	48.94	51.43	150.00	0.00	0.00
7	8	1	66.61	68.11	250.00	0.00	0.00
8	9	1	38.87	34.15	150.00	0.00	0.00
 Gen	 Status Fu	nel Pg	 Qg	Pmin	Pmax	Qmin	Qmax
1	 1 UNDEF	'INED 72.86	6.79	10.00	350.00	-300.00	300.00
2	1 UNDEF	INED 114.07	-5.13	10.00	300.00	-300.00	300.00
3	1 UNDEF	INED 80.21	-23.47	10.00	270.00	-300.00	300.00
[Exa	GO INFO]: Fi	nalizing opf	low appli	cation.			

# Example using the Ipopt [7]solver:

\$ ./bin/opflow -opflow_solver IPOPT -opflow_model POWEF -opflow_initialisation ACPF -opflow_objective MIN_GENSE [ExaGO INFO]: Checkingoptions_file not passed [ExaGO INFO]: Creating OPFlow	TPOINT_DEVIATION
****************	******
This program contains Ipopt, a library for large-scale	nonlinear
optimization. Ipopt is released as open source code und	
Eclipse Public License (EPL).	
************	* * * * * * * * * * * * * * * * * * * *
This is Ipopt version 3.12.10, running with linear solv	ver ma27.
Number of nonzeros in equality constraint Jacobian:	125
Number of nonzeros in inequality constraint Jacobian.:	106
Number of nonzeros in Lagrangian Hessian:	159
Total number of variables:	32
variables with only lower bounds:	0
variables with lower and upper bounds:	24
variables with only upper bounds:	0
Total number of equality constraints:	24
Total number of inequality constraints	36

```
inequality constraints with only lower bounds:
                                                            15
   inequality constraints with lower and upper bounds:
                                                            21
       inequality constraints with only upper bounds:
                                                            0
iter objective inf_pr inf_du lq(mu) ||d|| lq(rq) alpha_du alpha_pr ls
0 5.9999880e-04 1.80e+00 2.60e-01 -1.0 0.00e+00 - 0.00e+00 0.00e+00 0
1 1.1156839e-02 1.31e+00 7.65e+01 -1.0 1.25e+00 - 9.33e-02 2.67e-01h 1
2 1.9076073e-02 1.26e+00 7.36e+01 -1.0 1.84e+00 - 3.08e-02 4.25e-02f 1
3 4.6334972e-02 1.17e+00 6.85e+01 -1.0 1.20e+00 - 1.51e-02 7.14e-02f 1
86 8.8262569e-02 3.01e-12 2.18e-03 -7.0 2.12e-05 - 1.00e+00 1.00e+00h 1
87 8.8262567e-02 3.55e-15 1.33e-10 -7.0 2.54e-09 - 1.00e+00 1.00e+00h 1
Number of Iterations...: 87
                              (scaled)
                                                       (unscaled)
Objective..... 8.8262566709882084e-02
                                                 8.8262566709882084e-02
Dual infeasibility..: 1.3274141855146966e-10
                                                 1.3274141855146966e-10
Constraint violation: 3.5527136788005009e-15
                                                 3.5527136788005009e-15
Complementarity....: 9.0929465340161350e-08 9.0929465340161350e-08
Overall NLP error...: 9.0929465340161350e-08
                                                 9.0929465340161350e-08
Number of objective function evaluations
                                                    = 120
Number of objective gradient evaluations
                                                   = 88
Number of equality constraint evaluations
                                                   = 120
Number of inequality constraint evaluations
                                                   = 120
Number of equality constraint Jacobian evaluations = 88
Number of inequality constraint Jacobian evaluations = 88
Number of Lagrangian Hessian evaluations
Total CPU secs in IPOPT (w/o function evaluations)
                                                   =
                                                           0.039
Total CPU secs in NLP function evaluations
                                                           0.011
EXIT: Optimal Solution Found.
[ExaGO INFO]: Finalizing opflow application.
```

#### Example using the HiOp [6, 5] solver on the CPU:

```
# Hiop options
# end of Hiop options
Using 1 MPI ranks.
Problem Summary
_____
Total number of variables: 24
     lower/upper/lower_and_upper bounds: 16 / 16 / 16
Total number of equality constraints: 18
Total number of inequality constraints: 18
    lower/upper/lower_and_upper bounds: 18 / 18 / 18
iter objective inf_pr inf_du lg(mu) alpha_du
                                                  alpha_pr linesrch
0 1.0318125e+04 1.800e+00 4.460e+03 -1.00 0.000e+00 0.000e+00
                                                                 -(-)
1 7.7139414e+03 1.167e+00 2.900e+03 -1.00 6.255e-01 3.496e-01
                                                                  1(s)
2 7.3644945e+03 1.075e+00 2.654e+03 -1.00 9.785e-03 8.498e-02
                                                                 1(s)
. . .
13 4.1444605e+03 3.989e-10 3.273e-08 -7.00 1.000e+00 1.000e+00 1(h)
Successfull termination.
Total time 0.003 sec
Hiop internal time: total 0.001 sec
                                        avg iter 0.000 sec
    internal total std dev across ranks 0.000 percent
Fcn/deriv time:total=0.001sec (obj=0.000 grad=0.000 cons=0.000 \
                              Jac=0.000 Hess=0.001)
    Fcn/deriv total std dev across ranks 0.000 percent
Fcn/deriv #: obj 14 grad 14 eq cons 14 ineq cons 14 eq Jac 14 ineq Jac 14
Total KKT time 0.001 sec
       update init 0.000sec update linsys 0.000 sec
                                                         fact 0.001 sec
       solve rhs-manip 0.000 sec triangular solve 0.000 sec
[ExaGO INFO]: Finalizing opflow application.
```

#### Example using HiOp [6, 5] solver on the GPU:

```
# Hiop options
# end of Hiop options
Using 1 MPI ranks.
Problem Summary
_____
Total number of variables: 476
     lower/upper/lower_and_upper bounds: 277 / 277 / 277
Total number of equality constraints: 400
Total number of inequality constraints: 0
    lower/upper/lower_and_upper bounds: 0 / 0 / 0
iter objective inf_pr inf_du
                               lg(mu) alpha_du
                                                  alpha_pr linesrch
0 2.6329021e+04 1.725e+00 2.321e+03 -1.00 0.000e+00 0.000e+00 -(-)
1 2.6227762e+04 1.723e+00 2.318e+03 -1.00 8.506e-04 1.026e-03 1(s)
2 2.6134620e+04 1.721e+00 2.314e+03 -1.00 1.697e-03 8.630e-04 1(s)
24 2.7557571e+04 3.921e-09 3.801e-07 -7.00 1.000e+00 1.000e+00 1(h)
Successfull termination.
Total time 4.555 sec
Hiop internal time: total 4.538 sec
                                        avg iter 0.189 sec
    internal total std dev across ranks 0.000 percent
Fcn/deriv time: total=0.015 sec (obj=0.001 grad=0.001 cons=0.003 \setminus
                                     Jac=0.004 Hess=0.006)
   Fcn/deriv total std dev across ranks 0.000 percent
Fcn/deriv #: obj 27 grad 25 eq cons 27 ineq cons 27 eq Jac 25 ineq Jac 25
Total KKT time 4.464 sec
       update init 0.004sec update linsys 0.046 sec
                                                         fact 4.327 sec
       solve rhs-manip 0.072 sec triangular solve 0.015 sec
[ExaGO INFO]: Finalizing opflow application.
```

# Chapter 4

# Multi-period optimal power flow (TCOPFLOW)

TCOPFLOW solves a full AC multi-period optimal power flow problem with the objective of minimizing the total cost over the given time horizon while adhering to constraints for each period and between consecutive time-periods (ramping constraints).

#### 4.1 Formulation

The multi-period optimal power flow problem is a series of optimal power flow problems coupled via temporal constraints. The generator real power deviation  $(p_{jt}^{\rm g} - p_{jt-\Delta t}^{\rm g})$  constrained within the ramp limits form the temporal constraints. An illustration of the temporal constraints is shown in Fig. 4.1 with four time steps. Each time-step t is coupled with its preceding time  $t - \Delta t$ , where  $\Delta t$  is the time-step where the objective is to find a least cost dispatch for the given time horizon.

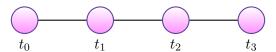


Figure 4.1: Multi-period optimal power flow example with four time-steps. The lines connecting the different time-periods denote the coupling between them.

In general form, the equations for multi-period optimal power flow are given by (4.1) – (4.5). TCOPFLOW solves to minimize the total generation cost  $\sum_{t=0}^{N_t-1} f(x_t)$  over the time horizon, where  $N_t$  is the number of time-steps. At each time-step, the equality constraints  $(g(x_t))$ , inequality  $h(x_t)$ , and the lower/upper limit  $(x^-, x^+)$  constraints need to be satisfied. Equation (4.5) represents the coupling between the consecutive time-steps. It is the most common form of coupling that limits the deviation of the real power generation at time t from its preceding time-step  $t - \Delta t$  to within its ramping capability  $\delta_t x$ .

$$\min \sum_{t=0}^{N_t - 1} f(x_t) \tag{4.1}$$

s.t

$$g(x_t) = 0, t \in [0, N_t - 1] (4.2)$$

$$h(x_t) \le 0,$$
  $t \in [0, N_t - 1]$  (4.3)

$$x^{-} \le x_t \le x^{+}, \qquad t \in [0, N_t - 1]$$
 (4.4)

$$-\delta_t x \le x_t - x_{t-\Delta t} \le \delta_t x, \qquad t \in [1, N_t - 1] \tag{4.5}$$

#### 4.2 Solvers

TCOPFLOW solves the coupled multi-period using the solver Ipopt [7]. Currently, ExaGO only supports solving TCOPFLOW using Ipopt [7], which is also the default solver. Solving TCOPFLOW is supported only on a single process.

#### 4.3 Input and Output

- **Network file:** The network file describing the network details. Only MATPOWER format files are currently supported.
- Load data: One file for load real power and one for reactive power. The files need to be in CSV format. An example of the format for the 9-bus case is here.
- Wind generation: The wind generation time-series described in CSV format. See an example of the format here.

If the load data and/or wind generation profiles are not set then a flat profile is assumed, i.e., the load and wind generation for all hours is constant.

The TCOPFLOW output is saved to a directory named tcopflowout. This directory contains  $N_t$  files, one for each time-step, in MATPOWER data file format.

## 4.4 Usage

```
./tcopflow -netfile <netfilename> -tcopflow_pload_profile <ploadprofile> \
-tcopflow_windgen_profile <windgenprofile> <tcopflowoptions> \
-tcopflow_qload_profile <qloadprofile>
```

## 4.5 Options

See table 4.1. In addition, all OPFLOW options given in Table 3.4 can be used.

Table 4.1: TCOPFLOW options

Option	Meaning	Values (Default value)
-netfile	Network file name	string < 4096 characters
		$(case9mod\_gen3\_wind.m)$
-save_output	Save output to file	0 or 1 (0)
-tcopflow_pload_profile	Real power load profile	string < 4096 characters
		(load_P.csv)
-tcopflow_qload_profile	Reactive power load profile	string < 4096 characters
		$(load\_Q.csv)$
-tcopflow_windgen_profile	Wind generation profile	string < 4096 characters
		(case9/scenarios_9bus.csv)
-tcopflow_dT	Length of time-step (min-	double $(5.0)$
	utes)	
-tcopflow_duration	Total duration (hours)	double $(0.5)$
-tcopflow_iscoupling	Coupling between time	0 or 1 (0)
	steps (ramp constraints)	

#### 4.6 Examples

Some TCOPFLOW example runs are provided with some sample output. Options are the default options provided in table 3.4 and 4.1 unless otherwise specified. Sample output is generated by running examples from the installation directory.

Example using Ipopt [7] solver:

```
$ ./bin/tcopflow -tcopflow_iscoupling 1 -print_output
[ExaGO INFO]: -- Checking ... -options_file not passed exists: no
TCOPFLOW: Application created
TCOPFLOW: Duration = 0.500000 hours, timestep = 5.000000 minutes, \
                                 number of time-steps = 7
TCOPFLOW: Using IPOPT solver
TCOPFLOW: Setup completed
Setting: "0" is not a valid setting for Option: tol.
Check the option documentation.
### tol (Real Number) ###
Category: Convergence
Description: Desired convergence tolerance (relative).
0 < (1e-08) <= +inf
*****************
This program contains Ipopt...
****************
This is Ipopt version 3.12.10, running with linear solver ma27.
Number of nonzeros in equality constraint Jacobian...:
                                                      770
Number of nonzeros in inequality constraint Jacobian.:
                                                      610
Number of nonzeros in Lagrangian Hessian....:
                                                      630
```

```
Total number of variables.....
                                                       161
                   variables with only lower bounds:
                                                        0
              variables with lower and upper bounds:
                                                       105
                   variables with only upper bounds:
                                                        0
Total number of equality constraints....:
                                                       126
Total number of inequality constraints....:
                                                       186
       inequality constraints with only lower bounds:
                                                        42
  inequality constraints with lower and upper bounds:
                                                       144
       inequality constraints with only upper bounds:
                                                         0
iter objective inf_pr inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr ls
0 5.2094875e+04 1.80e+00 5.02e+01 -1.0 0.00e+00 - 0.00e+00 0.00e+00
1 3.6277420e+04 1.40e+00 1.06e+02 -1.0 2.05e+00 - 1.88e-01 2.20e-01f 1
33 2.0843675e+04 1.13e-14 3.32e-09 -8.6 1.76e-08 - 1.00e+00 1.00e+00h 1
Number of Iterations...: 33
                                (scaled)
                                                        (unscaled)
Objective..... 4.6734697780313832e+02 2.0843675210019970e+04
Dual infeasibility..: 3.3166774138882895e-09
                                              1.4792381265941772e-07
Constraint violation: 1.1296519275560968e-14 1.1296519275560968e-14 Complementarity....: 4.9900570925632554e-09 2.2255654632832121e-07
Overall NLP error...: 4.9900570925632554e-09
                                             2.2255654632832121e-07
Number of objective function evaluations
                                               = 47
Number of objective gradient evaluations
                                                = 34
Number of equality constraint evaluations
Number of inequality constraint evaluations
                                                = 47
Number of equality constraint Jacobian evaluations = 34
Number of inequality constraint Jacobian evaluations = 34
Number of Lagrangian Hessian evaluations
                                                = 33
Total CPU secs in IPOPT (w/o function evaluations) =
                                                       0.025
Total CPU secs in NLP function evaluations
                                                       0.032
EXIT: Optimal Solution Found.
______
       Multi-Period Optimal Power Flow
_____
OPFLOW Model
                                 POWER BALANCE POLAR
Solver
                                 IPOPT
Duration (minutes)
                                 30.00
Time-step (minutes)
                                 5.00
Number of steps
Active power demand profile
                                datafiles/case9/load_P.csv
```

d gene d loss er imb ore li ber of ber of	eration s allowe balance ine flow f variab f equali f inequa	profile d allowed constra	ints raints straint	da NC NC 16 12						
_	nce stat e value	us			ONVERGED )843.68					
	Pd	Qd	Vm	Va	mu	 lt_Pmis	mult_Qn	 nis		
	0.00	0.00	1.040	0.000	) 2	 063.35	-0.00			
	0.00	0.00	1.025	4.553	3 2	013.40	0.00			
	0.00	0.00	1.025	2.955	5 2	017.53	0.00			
	0.00	0.00	1.037	-2.174	1 2	063.59	-0.74			
						074.28				
						092.22	4.26			
		0.00					3.85			
1	100.00	35.00	1.022	-1.667	7 2	035.30	7.65			
	0.00	0.00	1.036	0.264	1 2	018.00	3.11			
————— m ————	То	Stat	 us	Sft	Stf	 Slim	mult_Sf	mult_St		
	4	1					-0.00			
	7	1	11	1.80	112.67	250.00	-0.00	-0.00		
	9	1	8	36.66	87.56	300.00	-0.00	-0.00		
	5	1		28.34	34.62	250.00	-0.00	-0.00		
	6	1		12.81	45.47	250.00	-0.00	-0.00		
	7	1		17.86	51.21	250.00	-0.00	-0.00		
	9	1		19.48	52.48	150.00	-0.00	-0.00		
	8	1		3.89	65.25	250.00	-0.00	-0.00		
	8 9	1 1		63.89 11.64	65.25 36.64	250.00 150.00	-0.00 -0.00	- C		

Gen	Status	Fuel	Pg	Qg	Pmin	Pmax	Qmin	Qmax
1	1	COAL	71.06	5.89	10.00	350.00	-300.00	300.00
2	1	COAL	111.38	-9.71	10.00	300.00	-300.00	300.00
3	1	WIND	85.00	-16.87	0.00	85.00	-300.00	300.00
[Ex	aGO INFO	]: Final	Lizing tco	opflow app	lication.			

# Chapter 5

# Security-constrained optimal power flow (SCOPFLOW)

SCOPFLOW solves a contingency-constrained optimal power flow problem. The problem is set up as a two-stage optimization problem where the first-stage (base-case) represents the normal operation of the grid and the second-stage comprises  $N_c$  contingency scenarios. Each contingency scenario can be single or multi-period.

#### 5.1 Formulation

#### 5.1.1 Single-period

The contingency-constrained optimal power flow (popularly termed as security-constrained optimal power flow (SCOPF) in power system parlance) attempts to find a least cost dispatch for the base case (or no contingency) while ensuring that if any of contingencies do occur then the system will be secure. This is illustrated in Fig. 5.1 for a SCOPF with a base-case  $c_0$  and three contingencies.

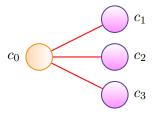


Figure 5.1: Contingency constrained optimal power flow example with three contingencies.  $c_0$  represents the base case (or no contingency case).  $c_1$ ,  $c_2$ ,  $c_3$  are the three contingency cases. Each of the contingency states is coupled with the base-case through ramping constraints (denoted by red lines)

In general form, the equations for contingency-constrained optimal power flow are given by (5.1) – (5.5). This is a two-stage stochastic optimization problem where the first stage is the base case  $c_0$  and each of the contingency states  $c_i$ ,  $i \in [1, N_c - 1]$  are second-stage subproblems. SCOPFLOW aims to minimize the objective  $\sum_{c=0}^{N_c-1} f(x_c)$ , while adhering to the equality  $g(x_c)$ , inequality  $h(x_c)$ , and the lower/upper bound  $(x^-, x^+)$  constraints. Equation (5.5) represents the coupling between the base-case and each of the contingency states  $c_i$ . Equation (5.5) is the most typical form of coupling that limits the deviation of the contingency variables  $x_c$  from the base  $x_0$  to within  $\delta_c x$ . An example of this constraint could be the allowed real power output deviation for the generators constrained by their ramp limit.

$$\min f(x_0) \tag{5.1}$$

s.t.

$$g(x_c) = 0, c \in [0, N_c - 1] (5.2)$$

$$h(x_c) \le 0,$$
  $c \in [0, N_c - 1]$  (5.3)

$$x^{-} \le x_c \le x^{+}, \qquad c \in [0, N_c - 1]$$
 (5.4)

$$-\delta_c x < x_c - x_0 < \delta_c x, \qquad c \in [1, N_c] \tag{5.5}$$

#### 5.1.2 Multiperiod

In the multi-period version, each contingency is comprised of multiple time-periods. The multiple periods have variables and constraints as described in chapter 4. An example of multi-contingency multiperiod optimal power flow is illustrated in Fig. 5.2 with two contingencies  $c_0$  and  $c_1$ . Here,  $c_0$  is the case with no contingencies, i.e., the base-case. Each contingency is multi-period with four time-periods. Each time-step is coupled with its adjacent one through ramping constraints. We assume that the contingency is incident at the first time-step, i.e. at  $t_0$ . This results in the coupling between the contingency cases  $c_i$ ,  $i \in [1, N_c - 1]$  and the base-case  $c_0$  only at time-step  $t_0$  as shown in Fig. 5.2.

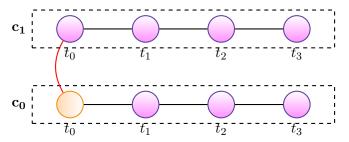


Figure 5.2: Multi-period contingency constrained optimal power flow example with two contingencies  $c_0$  and  $c_1$ , each with four time-periods  $t_0$ ,  $t_1$ ,  $t_2$ ,  $t_3$ . State  $c_0$ ,  $t_0$  represent the base case (no contingency) case. We assume that any contingency is incident at the first time-step, i.e., at  $t_0$ . The contingency states  $c_1$ ,  $t_0$  is coupled with the no-contingency state  $c_0$ ,  $t_0$  at time  $t_0$ . The red line denotes the coupling between the contingency.

The overall objective of this contingency-constrained multi-period optimal power flow is to find a secure dispatch for base-case  $c_0$  while adhering to contingency and temporal constraints. The general formulation of this problem is given in Eqs. (5.6) - (5.12).

$$\min \sum_{c=0}^{N_c-1} \sum_{t=0}^{N_t-1} f(x_{c,t}) \tag{5.6}$$

s.t.

$$g(x_{c,t}) = 0,$$
  $c \in [0, N_c - 1], t \in [0, N_t - 1]$  (5.7)

$$h(x_{c,t}) \le 0,$$
  $c \in [0, N_c - 1], t \in [0, N_t - 1]$  (5.8)

$$x^{-} \le x_{c,t} \le x^{+},$$
  $c \in [0, N_c - 1], t \in [0, N_t - 1]$  (5.9)

$$-\delta_t x \le x_{c,t} - x_{c,t-\Delta t} \le \delta_t x, \qquad c \in [0, N_c - 1], t \in [1, N_t - 1]$$
 (5.10)

$$-\delta_c x \le x_{c,0} - x_{0,0} \le \delta_c x, \qquad c \in [1, N_c - 1] \tag{5.11}$$

(5.12)

In this formulation, the objective is to reduce the cost for the base-case time horizon, where  $f(x_{0,t})$  is the objective cost for contingency  $c_0$  at time t. Equation (5.12) represents the coupling between the base case  $c_0$  and each contingency  $c_i$  at time-step  $t_0$ . We use a simple box constraint  $\delta_c x$  to restrict the deviation of decision variables  $x_{c,0}$  from the base-case  $x_{0,0}$ . The bound  $\delta_c x$  could represent here, for example, the allowable reserve for each generator.

#### 5.2 Solvers

SCOPFLOW can be solved with Ipopt [7]. If one wants to solve each contingency independently, i.e., without any coupling constraints then use the *EMPAR* solver. *EMPAR* distributes the contingencies to different processes when executed in parallel.

#### 5.3 Input and Output

To execute SCOPFLOW, the following files are required:

- **Network file:** The network file describing the network details. Only MATPOWER format files are currently supported.
- Contingency file: The file describing the contingencies. Contingencies can be single or multiple outages. The contingency file needs to be described in PTI format.

If the multi-period option is chosen, then additional files describing the load and wind generation can be (optionally) set.

- Load data: One file for load real power and one for reactive power. The files need to be in CSV format. An example of the format for the 9-bus case is here.
- Wind generation: The wind generation time-series described in CSV format. See an example of the format here.

The SCOPFLOW output is saved to a directory named *scopflowout*. This directory contains  $N_c$  files to save the solution for each contingency in MATPOWER datafile format. Each file has the name  $cont\_xx$  where xx is the contingency number.

If the multi-period option is chosen then  $N_c$  subdirectories are created (one for each contingency), and each subdirectory contains  $N_t$  output files, one for each time-period. The subdirectories have the naming convention  $cont_{-}xx$  and the output file are named as  $t_{-}yy$  where yy is the time-step number.

#### 5.4 Usage

```
./scopflow -netfile <netfilename> -ctgcfile <ctgcfilename> \ <scopflowoptions> [-scopflow_enable_multiperiod 1]
```

## 5.5 Options

See table 5.1. In addition, all OPFLOW options in Table 3.4 and TCOPFLOW options in Table 4.1 can be used.

Option	Meaning	Values (Default value)
-netfile	Network file name	string < 4096 characters
		$(case9mod\_gen3\_wind.m)$
-ctgcfile	Contingency file name	string < 4096 characters
		(case9.cont)
-save_output	Save output to directory	0 or 1 (0)
-scopflow_solver	Set solver for scopflow	IPOPT or EMPAR
-scopflow_Nc	Number of contingencies	int (0. Passing -1 results in all
		contingencies in the file used)
-scopflow_mode	Operation mode: Preventive or	0 or 1 (0)
	corrective	
-scopflow_iscoupling	Coupling between first and second	0 or 1 (0)
	stage	
-scopflow_enable_multiperiod	Multi-period SCOPFLOW	0 or 1 (0)
-scopflow_pload_profile	Real power load profile	string (load_P.csv)
-scopflow_qload_profile	Reactive power load profile	string (load_Q.csv)
-scopflow_windgenprofile	Wind generation profile	string (case9/scenarios_9bus.csv)
-scopflow_dT	Length of time-step (minutes)	double $(5.0)$
-scopflow_duration	Total duration (hours)	double $(0.5)$

Table 5.1: SCOPFLOW options

Depending on the *mode*, SCOPFLOW can either be *preventive* (mode = 0) or *corrective* (mode = 1). In the preventive mode, the PV and PQ generator real power is fixed to its corresponding base-case values. The generators at the reference bus pick up any make-up power required for the contingency. The corrective mode allows deviation of the PV and PQ generator real power from the base-case dispatch constrained by its 30-min. ramp rate capability.

-scopflow\_enable\_multiperiod 1 must be used in order to enable any of the options listed in table 5.1 that are below -scopflow\_enable\_multiperiod.

# 5.6 Examples

Some SCOPFLOW example runs are provided with some sample output. Options are the default options given in table 3.4, 4.1 and 5.1 unless otherwise specified. Sample output is generated by running examples in the installation directory.

Example using the Ipopt [7] solver:

```
$ ./bin/scopflow -scopflow_solver IPOPT -tcopflow_iscoupling 1 \
```

```
-scopflow_iscoupling 1 -print_output
[ExaGO INFO]: -- Checking ... -options_file not passed exists: no
SCOPFLOW: Application created
SCOPFLOW running with 2 contingencies (base case + 1 contingencies)
Rank 0 has 2 contingencies, range [0 -- 2]
SCOPFLOW: Using IPOPT solver
SCOPFLOW: Setup completed
*************
This program contains Ipopt, a library for large-scale nonlinear
optimization. Ipopt is released as open source code under the Eclipse
Public License (EPL).
*****************
This is Ipopt version 3.12.10, running with linear solver ma27.
Number of nonzeros in equality constraint Jacobian...:
                                                      231
Number of nonzeros in inequality constraint Jacobian .:
                                                      182
Number of nonzeros in Lagrangian Hessian....:
                                                      249
Total number of variables....:
                                                       55
                   variables with only lower bounds:
                                                        0
                                                       39
              variables with lower and upper bounds:
                   variables with only upper bounds:
                                                       0
Total number of equality constraints....:
                                                       44
Total number of inequality constraints....:
                                                       59
       inequality constraints with only lower bounds:
                                                       21
  inequality constraints with lower and upper bounds:
                                                       38
       inequality constraints with only upper bounds:
iter objective inf_pr inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr
0 7.4421250e+03 1.80e+00 1.69e+01 -1.0 0.00e+00 - 0.00e+00 0.00e+00
1 6.9108233e+03 1.65e+00 1.56e+01 -1.0 1.24e+00 - 4.46e-02 8.38e-02f
                                                                 1
2 5.9991119e+03 1.30e+00 5.10e+01 -1.0 7.41e-01 - 7.07e-02 2.08e-01f
43 3.0556401e+03 1.07e-14 3.95e-10 -8.6 2.79e-09 - 1.00e+00 1.00e+00f
Number of Iterations...: 43
                                (scaled)
                                                      (unscaled)
                    6.8512110967619464e+01
Objective....:
                                             3.0556401491558281e+03
Dual infeasibility..: 3.9453703519053556e-10
                                             1.7596351769497887e-08
Constraint violation:
                    1.0658141036401503e-14
                                             1.0658141036401503e-14
Complementarity....: 5.0027986112663857e-09
                                             2.2312481806248081e-07
Overall NLP error...:
                     5.0027986112663857e-09
                                             2.2312481806248081e-07
```

Number	of	objective function evaluations	=	59	
Number	of	objective gradient evaluations	=	44	
Number	of	equality constraint evaluations	=	59	
Number	of	inequality constraint evaluations	=	59	
Number	of	equality constraint Jacobian evaluations	=	44	
Number	of	inequality constraint Jacobian evaluations	=	44	
Number	of	Lagrangian Hessian evaluations	=	43	
Total (	CPU	secs in IPOPT (w/o function evaluations)	=		0.019
Total (	CPU	secs in NLP function evaluations	=		0.009
1					

#### EXIT: Optimal Solution Found.

\_\_\_\_\_

#### Security-Constrained Optimal Power Flow

\_\_\_\_\_

OPFLOW Formulation	POWER_BALANCE_POLAR
Solver	IPOPT
Initialization	MIDPOINT
Number of contingencies	1
Load loss allowed	NO
Power imbalance allowed	NO
Ignore line flow constraints	NO
Number of variables	57
Number of equality constraints	42
Number of inequality constraints	58
Number of coupling constraints	3
I	

Convergence status CONVERGED Objective value 3055.64

Pd Qd Vm Va mult\_Pmis mult\_Qmis \_\_\_\_\_ 0.00 0.00 1.040 0.000 2158.80 -0.00 0.00 0.00 1.025 4.572 2108.44 0.00 0.00 0.00 1.025 1.912 3 2118.81 0.00 0.00 0.00 1.038 -2.306 2159.07 -0.6175.00 30.00 1.026 -3.550 5 2170.87 2.93 6 90.00 30.00 1.023 -4.551 2191.34 4.30 7 0.00 0.00 1.033 0.613 2109.20 4.12 100.00 35.00 1.022 -2.070 2133.92 7.92 0.00 0.00 1.036 -0.4622119.23 2.84 From

 From
 To
 Status
 Sft
 Stf
 Slim
 mult\_Sf
 mult\_St

 1
 4
 1
 75.61
 75.44
 380.00
 -0.00
 -0.00

2	7		1	117.34	118.26	250.00	-0.00	-0.00	
3	9		1	76.90	77.68	300.00	-0.00	-0.00	
4	5		1	28.64	34.88	250.00	-0.00	-0.00	
4	6		1	46.84	48.91	250.00	-0.00	-0.00	
5	7		1	47.56	50.96	250.00	-0.00	-0.00	
6	9		1	45.96	48.57	150.00	-0.00	-0.00	
7	8		1	69.77	70.78	250.00	-0.00	-0.00	
8	9		1	37.09	30.76	150.00	-0.00	-0.00	
Gen	Status	Fuel	Pg	Qg 	Pmin	Pmax	Qmin	Qmax	
1	1	COAL	75.40	5.62	10.00	350.00	-300.00	300.00	
2	1	COAL	116.97	-9.31	10.00	300.00	-300.00	300.00	
3	1	WIND	75.00	-16.97	0.00	75.00	-300.00	300.00	
[ExaGC	ExaGO INFO]: Finalizing scopflow application								

#### Example using the IPOPT solver with multiperiod enabled:

```
$./bin/scopflow -scopflow_solver IPOPT -scopflow_enable_multiperiod 1 \
-tcopflow_iscoupling 1 -scopflow_iscoupling 1
[ExaGO INFO]: -- Checking ... -options_file not passed
                                                         exists: no
SCOPFLOW: Application created
SCOPFLOW running with 2 contingencies (base case + 1 contingencies)
Rank 0 has 2 contingencies, range [0 -- 2]
SCOPFLOW: Using IPOPT solver
TCOPFLOW: Application created
TCOPFLOW: Duration = 0.166667 hours, timestep = 5.000000 minutes, \
         number of time-steps = 3
TCOPFLOW: Using IPOPT solver
TCOPFLOW: Setup completed
TCOPFLOW: Application created
TCOPFLOW: Duration = 0.166667 hours, timestep = 5.000000 minutes, \
         number of time-steps = 3
TCOPFLOW: Using IPOPT solver
TCOPFLOW: Setup completed
SCOPFLOW: Setup completed
******************
This program contains Ipopt, a library for large-scale nonlinear
optimization. Ipopt is released as open source code under the
Eclipse Public License (EPL).
*****************
This is Ipopt version 3.12.10, running with linear solver ma27.
Number of nonzeros in equality constraint Jacobian...:
                                                       640
Number of nonzeros in inequality constraint Jacobian.:
                                                       494
```

```
Number of nonzeros in Lagrangian Hessian....:
                                                         540
Total number of variables....:
                                                         138
                    variables with only lower bounds:
                                                           0
               variables with lower and upper bounds:
                                                          90
                    variables with only upper bounds:
                                                           0
Total number of equality constraints....:
                                                         110
Total number of inequality constraints....:
                                                         151
       inequality constraints with only lower bounds:
                                                          36
   inequality constraints with lower and upper bounds:
                                                         115
       inequality constraints with only upper bounds:
                                                           0
iter objective inf pr inf du lq(mu) ||d|| lq(rq) alpha du alpha pr
0 4.4652750e+04 1.80e+00 5.01e+01 -1.0 0.00e+00 - 0.00e+00 0.00e+00
1 3.1127309e+04 1.41e+00 1.01e+02 -1.0 1.97e+00 - 1.87e-01 2.18e-01f
55 1.7423413e+04 9.91e-15 8.48e-10 -8.6 9.66e-09 - 1.00e+00 1.00e+00h 1
Number of Iterations....: 55
                                 (scaled)
                                                         (unscaled)
Objective..... 3.9065947675497836e+02
                                                1.7423412663272036e+04
Dual infeasibility..: 8.4833681960093011e-10
                                                3.7835822154201486e-08
Constraint violation: 9.9087404947795221e-15
                                                9.9087404947795221e-15
Complementarity....: 3.3875752035319637e-09
                                                1.5108585407752560e-07
Overall NLP error...: 3.3875752035319637e-09
                                               1.5108585407752560e-07
Number of objective function evaluations
                                                  = 91
Number of objective gradient evaluations
                                                  = 56
Number of equality constraint evaluations
                                                  = 91
Number of inequality constraint evaluations
                                                  = 91
Number of equality constraint Jacobian evaluations
                                                  = 56
Number of inequality constraint Jacobian evaluations = 56
Number of Lagrangian Hessian evaluations
                                                  = 55
Total CPU secs in IPOPT (w/o function evaluations)
                                                         0.045
Total CPU secs in NLP function evaluations
                                                         0.052
EXIT: Optimal Solution Found.
[ExaGO INFO]: Finalizing scopflow application.
```

#### Example using the *EMPAR* solver with multiperiod enabled:

```
$ ./bin/scopflow -scopflow_solver EMPAR -scopflow_enable_multiperiod 1 \
-tcopflow_iscoupling 1 -scopflow_iscoupling 1

[ExaGO INFO]: -- Checking ... -options_file not passed exists: no
SCOPFLOW: Application created
SCOPFLOW running with 2 contingencies (base case + 1 contingencies)
```

```
Rank 0 has 2 contingencies, range [0 -- 2]
SCOPFLOW: Using EMPAR solver
TCOPFLOW: Application created
TCOPFLOW: Duration = 0.166667 hours, timestep = 5.000000 minutes,
         number of time-steps = 3
TCOPFLOW: Using IPOPT solver
TCOPFLOW: Setup completed
TCOPFLOW: Application created
TCOPFLOW: Duration = 0.166667 hours, timestep = 5.000000 minutes,
         number of time-steps = 3
TCOPFLOW: Using IPOPT solver
TCOPFLOW: Setup completed
SCOPFLOW: Setup completed
Setting: "0" is not a valid setting for Option: tol.
Check the option documentation.
### tol (Real Number) ###
Category: Convergence
Description: Desired convergence tolerance (relative).
0 < (1e-08) <= +inf
************
This program contains Ipopt, a library for large-scale nonlinear
optimization. Ipopt is released as open source code under
the Eclipse Public License (EPL).
*****************
This is Ipopt version 3.12.10, running with linear solver ma27.
Number of nonzeros in equality constraint Jacobian...:
                                                          330
Number of nonzeros in inequality constraint Jacobian.:
                                                          258
Number of nonzeros in Lagrangian Hessian....:
                                                          270
Total number of variables....:
                                                           69
                    variables with only lower bounds:
                                                            0
               variables with lower and upper bounds:
                                                           45
                    variables with only upper bounds:
                                                            0
Total number of equality constraints....:
                                                           54
Total number of inequality constraints....:
                                                           78
        inequality constraints with only lower bounds:
                                                           18
   inequality constraints with lower and upper bounds:
                                                           60
       inequality constraints with only upper bounds:
                                                            0
iter objective inf_pr inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr ls
0 \quad 2.2326375 \text{e} + 04 \quad 1.80 \text{e} + 00 \quad 5.02 \text{e} + 01 \quad -1.0 \quad 0.00 \text{e} + 00 \quad -0.00 \text{e} + 00 \quad 0.00 \text{e} + 00
1 1.5533385e+04 1.40e+00 1.02e+02 -1.0 1.95e+00 - 1.88e-01 2.20e-01f
. . .
```

```
8.6735595e+03 7.99e-15 3.70e-09 -8.6 1.93e-08 - 1.00e+00 1.00e+00h 1
Number of Iterations...: 32
                                   (scaled)
                                                            (unscaled)
Objective....: 1.9447442817603635e+02
                                                  8.6735594966512217e+03
Dual infeasibility..: 3.6983296779767669e-09
                                                 1.6494550363776382e-07
Constraint violation: 7.9936057773011271e-15
                                                  7.9936057773011271e-15
Complementarity....: 4.9913149587308981e-09 2.2261264715939807e-07
Overall NLP error...: 4.9913149587308981e-09 2.2261264715939807e-07
Number of objective function evaluations
                                                    = 40
Number of objective gradient evaluations
                                                   = 33
Number of equality constraint evaluations
                                                    = 40
Number of inequality constraint evaluations
                                                    = 40
Number of equality constraint Jacobian evaluations = 33
Number of inequality constraint Jacobian evaluations = 33
Number of Lagrangian Hessian evaluations
                                                    = 32
Total CPU secs in IPOPT (w/o function evaluations)
                                                    =
                                                            0.019
Total CPU secs in NLP function evaluations
                                                            0.014
EXIT: Optimal Solution Found.
Setting: "0" is not a valid setting for Option: tol.
Check the option documentation.
### tol (Real Number) ###
Category: Convergence
Description: Desired convergence tolerance (relative).
0 < (1e-08) <= +inf
This is Ipopt version 3.12.10, running with linear solver ma27.
Number of nonzeros in equality constraint Jacobian...:
                                                            306
Number of nonzeros in inequality constraint Jacobian.:
                                                            234
Number of nonzeros in Lagrangian Hessian....:
                                                            270
Total number of variables....:
                                                             69
                     variables with only lower bounds:
                                                             0
                variables with lower and upper bounds:
                                                             45
                    variables with only upper bounds:
                                                             0
Total number of equality constraints....:
                                                             54
Total number of inequality constraints....:
                                                             72
        inequality constraints with only lower bounds:
                                                             18
   inequality constraints with lower and upper bounds:
                                                             54
        inequality constraints with only upper bounds:
                                                             0
iter objective inf_pr inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr ls
```

```
0 \quad 2.2326375e + 04 \quad 1.80e + 00 \quad 4.89e + 01 \quad -1.0 \quad 0.00e + 00 \quad -0.00e + 00 \quad 0.00e + 00
1 1.5535979e+04 1.41e+00 1.01e+02 -1.0 1.97e+00 - 1.87e-01 2.18e-01f 1
51 8.7491131e+03 8.88e-15 4.80e-09 -8.6 1.07e-08 - 1.00e+00 1.00e+00h 1
Number of Iterations....: 51
                                (scaled)
                                                       (unscaled)
Objective..... 1.9616845456090007e+02
                                                  8.7491130734161434e+03
                                                  2.1416475174532407e-07
Dual infeasibility..: 4.8019002633480731e-09
Constraint violation: 8.8817841970012523e-15
                                                  8.8817841970012523e-15
Complementarity....: 4.9761717933322852e-09 2.2193726198261992e-07
Overall NLP error...: 4.9761717933322852e-09 2.2193726198261992e-07
Number of objective function evaluations
                                                     = 54
Number of objective gradient evaluations
                                                     = 52
Number of equality constraint evaluations
                                                     = 54
Number of inequality constraint evaluations
                                                     = 54
Number of equality constraint Jacobian evaluations = 52
Number of inequality constraint Jacobian evaluations = 52
Number of Lagrangian Hessian evaluations
                                                     = 51
Total CPU secs in IPOPT (w/o function evaluations) =
                                                            0.024
Total CPU secs in NLP function evaluations
                                                     =
                                                            0.016
EXIT: Optimal Solution Found.
[ExaGO INFO]: Finalizing scopflow application.
```

## Chapter 6

# Stochastic optimal power flow (SOPFLOW)

SOPFLOW solves a stochastic security-constrained multi-period optimal power flow problem. The problem is set up as a two-stage optimization problem where the first-stage (base-case) represents the normal operation of the grid (or the most likely forecast) and the second-stage comprises  $N_s$  scenarios of forecast deviation. Each scenario can have multiple contingencies and each contingency can be multi-period. Thus, depending on the options selected, each stochastic scenario can be

- Single-period, no contingencies
- Single-period, with contingencies
- Multi-period, with contingencies

#### 6.1 Formulation

An illustration of SOPFLOW is shown in Fig. 6.1 for a case with two scenarios  $s_0$  and  $s_1$ . Each scenario has two contingencies  $c_0$ ,  $c_1$ , and each contingency has four time-periods.

The formulation for the stochastic security-constrained multi-period optimal power flow is given in (6.1) - (6.7). In this formulation, the objective is to reduce the expected cost, where  $f(x_{s,c,t})$  is the cost for scenario s with contingency c at time t.  $\pi_s$  is the probability of scenario s.

$$\min \sum_{s=0}^{N_s-1} \pi_s \sum_{c=0}^{N_c-1} \sum_{t=0}^{N_t-1} f(x_{s,c,t})$$
(6.1)

s.t.

$$g(x_{s,c,t}) = 0,$$
  $s \in [1, N_s - 1], c \in [0, N_c - 1], t \in [0, N_t - 1]$  (6.2)

$$h(x_{s,c,t}) \le 0,$$
  $s \in [1, N_s - 1], c \in [0, N_c - 1], t \in [0, N_t - 1]$  (6.3)

$$x^{-} \le x_{s,c,t} \le x^{+},$$
  $s \in [1, N_s - 1], c \in [0, N_c - 1], t \in [0, N_t - 1]$  (6.4)

$$-\delta_t x \le x_{s,c,t} - x_{s,c,t-\Delta t} \le \delta_t x, \qquad s \in [1, N_s - 1], c \in [0, N_c - 1], t \in [1, N_t - 1]$$
(6.5)

$$-\delta_c x \le x_{s,c,0} - x_{s,0,0} \le \delta_c x, \qquad s \in [1, N_s - 1], c \in [1, N_c - 1] \tag{6.6}$$

$$-\delta_s x \le x_{s,0,0} - x_{0,0,0} \le \delta_s x, \qquad s \in [1, N_s - 1] \tag{6.7}$$

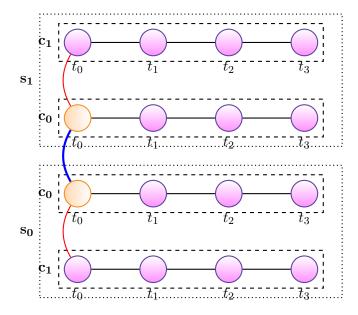


Figure 6.1: Stochastic multi-period contingency constrained example with two scenarios  $s_0$  and  $s_1$ . Each scenario has two contingencies  $c_0$  and  $c_1$  and each contingency consists of four time-periods  $t_0$ ,  $t_1$ ,  $t_2$ ,  $t_3$ . State  $s_0$ ,  $c_0$ ,  $t_0$  represents the base case (no contingency) case for the two scenarios. We assume that any contingency is incident at the first time-step, i.e., at  $t_0$ . Thus, the contingency states  $c_1$ ,  $t_0$  is coupled with the no-contingency state  $c_0$ ,  $t_0$  at time  $t_0$  for both the scenarios. The red line denotes the coupling between the contingency and the no-contingency states. The blue line denotes the coupling between the scenarios

The modeling details used for an optimial power flow problem are also used for a SOPFLOW problem, i.e., each of the circles shown in Fig. 6.1 has the modeling details of an optimal power flow problem (OPFLOW). Incorporating the probabilities  $\pi_s$  for each scenario is not implemented yet which leads to each scenario having an equal probability.

Currently, SOPFLOW uses wind power generation as the stochastic variables and each scenario is a realization of the power output from wind generators. A zero fuel cost is used for wind power generation to ensure wind generation would be the dispatched to the given target level (upper limit).

For contingencies, SOPFLOW supports generation and/or transmission outages. A contingency can have multiple outages, but, it should not cause any islanding. The coupling between the no-contingency and the contingency case for each scenario is also the difference in real power output  $(p_{jsct}^{\rm g} - p_{js0t}^{\rm g}, j \in J^{\rm gen})$  that must be within the 30 minute generator ramp rate.

For multi time-period, we use ramping constraints on the generator real power output between successive time steps.

SOPFLOW can be run in two modes: preventive and corrective. In the preventive mode, generator real power output is fixed to the base-case values for generators at PV bus(es). In this mode, the generators at the reference bus provide/absorb any deficit/surplus power. The corrective mode allows deviation of the PV and PQ generator real power from the base-case dispatch constrained by its 30-min. ramp rate capability. Note that the preventive/corrective mode is only applied at the first step  $t_0$ . In the successive time-steps, the generator dispatch is dictated by the previous step dispatch and the ramp limits.

#### 6.2 Solvers

SOPFLOW can be solved with Ipopt [7]. If one wants to solve each scenario independently, i.e., without any coupling constraints then use the *EMPAR* solver. *EMPAR* distributes the contingencies to different processes when executed in parallel.

### 6.3 Input and Output

The following files are needed for executing SOPFLOW.

- **Network file:** The network file describing the network details. Only MATPOWER format files are currently supported.
- Scenario file: SOPFLOW only supports reading wind generation scenarios in a CSV format. An example of this format for the 9-bus case is here.
- Contingency file: Contingencies can be specified via PTI format file as described in chapter 5. The option -sopflow\_enable\_multicontingency should be set for multi-contingency problems.
- Load data: One file for load real power and one for reactive power. The files need to be in CSV format. An example of the format for the 9-bus case is here.

The SOPFLOW output is saved to a directory named sopflowout. This directory contains  $N_s$  subdirectories to save the solution for each scenario. Each of these subdirectories contain  $N_c$  subdirectories, one for each contingency. Each contingency subdirectory has  $N_t$  MATPOWER format files to store the output for each time-period for the given contingency and scenario. The subdirectories have the directory name format  $scen_x$  where x is the scenario number,  $cont_y$  where y is the contingency number, and the output files have the file name format  $t_z$  where z is the time-step number.

## 6.4 Usage

```
./sopflow -netfile <netfilename> -scenfile <scenfilename> \
[-sopflow_enable_multicontingency 1] <sopflowoptions>
```

## 6.5 Options

See table 6.1. With multi-contingency SOPFLOW, all SCOPFLOW options given in Table 5.1 can be used to tune the contingencies. All OPFLOW options in Table 3.4, along with TCOPFLOW options in Table 4.1 can also be used. Multi-contingency SOPFLOW also allows the options listed after -sopflow\_enable\_multicontingency in Table 6.1 to be used.

Depending on the mode, SOPFLOW can either be preventive (mode = 0) or corrective (mode = 1). In the preventive mode, the PV and PQ generator real power is fixed to its corresponding base-case values. The generators at the reference bus pick up any make-up power required for the contingency. The corrective mode allows deviation of the PV and PQ generator real power from the base-case dispatch constrained by its 30-min. ramp rate capability.

Table 6.1: SOPFLOW options

Option	Meaning	Values (Default value)
-netfile	Network file name	string < 4096 characters
		(case9mod_gen3_wind.m)
-scenfile	Scenario file name	string < 4096 characters
		(case9/scenarios_9bus.csv)
-sopflow_mode	Operation mode: Preventive or	0 or 1 (0)
	corrective	
-sopflow_solver	Set solver for sopflow	IPOPT or EMPAR
-sopflow_enable_multicontingency	Multi-contingency SOPFLOW	0 or 1 (0)
-ctgcfile	Contingency file name	string (case9.cont)
-sopflow_Ns	Number of scenarios	int (Default 0. Passing -1 results
		in all scenarios in the file being
		used)

## 6.6 Examples

Some SOPFLOW example runs are provided with some sample output. Options are the default options given in Tables 3.4, 4.1, 5.1 and 6.1 unless otherwise specified. Sample output is generated by running examples in the installation directory.

Example using the Ipopt [7] solver:

```
$ ./bin/sopflow -print_output -sopflow_solver IPOPT
[ExaGO INFO]: -- Checking ... -options_file not passed
                                                   exists: no
SOPFLOW: Application created
Rank[0]: color = 0, ns = 2, sstart= 0, send=2
SOPFLOW running with 2 scenarios (base case + 1 scenarios)
Rank 0 scenario range [0 -- 2]
SOPFLOW: Using IPOPT solver
SOPFLOW: Setup completed
************
This program contains Ipopt, a library for large-scale nonlinear
optimization. Ipopt is released as open source code under the
Eclipse Public License (EPL).
*****************
This is Ipopt version 3.12.10, running with linear solver ma27.
Number of nonzeros in equality constraint Jacobian...:
                                                     2.2.2
Number of nonzeros in inequality constraint Jacobian .:
                                                     168
Number of nonzeros in Lagrangian Hessian....:
                                                     180
Total number of variables....:
                                                     46
                  variables with only lower bounds:
                                                      0
              variables with lower and upper bounds:
                                                     30
                  variables with only upper bounds:
                                                      0
Total number of equality constraints....:
                                                     37
```

```
Total number of inequality constraints.....
                                                          50
       inequality constraints with only lower bounds:
                                                         12
   inequality constraints with lower and upper bounds:
                                                          38
       inequality constraints with only upper bounds:
                                                          0
iter objective inf_pr inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr ls
0 1.4884250e+04 1.80e+00 5.02e+01 -1.0 0.00e+00 - 0.00e+00 0.00e+00
1 1.0889088e+04 1.45e+00 7.87e+01 -1.0 1.92e+00 - 1.69e-01 1.93e-01f
                                                                    1
2 1.0548318e+04 1.33e+00 7.28e+01 -1.0 6.01e-01 - 2.76e-02 8.30e-02f 1
29 5.9075224e+03 1.24e-14 2.96e-12 -7.0 4.91e-10 - 1.00e+00 1.00e+00h 1
Number of Iterations....: 29
                                                         (unscaled)
                                 (scaled)
Objective....: 1.3245566021454471e+02
                                               5.9075224455686939e+03
Dual infeasibility..: 2.9592247352450450e-12
                                              1.3198142319192901e-10
Constraint violation: 1.2434497875801753e-14 1.2434497875801753e-14 Complementarity....: 9.0910764708133484e-08 4.0546201059827535e-06
Overall NLP error...: 9.0910764708133484e-08 4.0546201059827535e-06
Number of objective function evaluations
                                                  = 33
Number of objective gradient evaluations
                                                 = 30
Number of equality constraint evaluations
                                                 = 33
Number of inequality constraint evaluations
                                                 = 33
Number of equality constraint Jacobian evaluations = 30
Number of inequality constraint Jacobian evaluations = 30
Number of Lagrangian Hessian evaluations
                                                 = 29
Total CPU secs in IPOPT (w/o function evaluations) =
                                                         0.006
Total CPU secs in NLP function evaluations
                                                         0.003
EXIT: Optimal Solution Found.
______
       Stochastic Optimal Power Flow
______
OPFLOW Formulation
                                  POWER_BALANCE_POLAR
Solver
                                  IPOPT
Initialization
                                  MIDPOINT
Number of scenarios
                                  1
Load loss allowed
                                  NO
Power imbalance allowed
                                  NO
Ignore line flow constraints
                                  NO
Number of variables
                                  48
Number of equality constraints
                                  \cap
Number of inequality constraints
                                  0
```

Numbe	er of coup	ling co	nstrain	ī.s	0				
	ergence st					NVERGED			
Obje	ctive valu	e			59	07.52			
Bus	Pd	Qd	. Vi	n	 Va	mul	 t_Pmis	mult_Qm	nis
1	0.00	0.0	0 1.0	40	0.000	22	 18.76	-0.00	
2	0.00	0.0	0 1.02	25	4.156	21	70.20	0.00	
3	0.00	0.0	0 1.02	25	1.685	21	80.00	0.00	
4	0.00	0.0			-2.389	22	19.06	-0.56	
5	75.00				-3.714		31.93	2.96	
6	90.00				-4.683		52.92	4.40	
7	0.00				0.291		70.95	4.07	
8	100.00				-2.352		96.02	8.01	
9	0.00	0.0	0 1.03	36	-0.689	21	80.43	2.83	
From	То	 S	tatus	 S	ft	Stf	Slim	mult_Sf	mult_St
1	4		1	78	.31	78 <b>.</b> 15	380.00	-0.00	-0.00
2	7		1	114	.58	115.48	250.00	-0.00	-0.00
3	9		1	76	.90	77.69	300.00	-0.00	-0.00
4	5		1	30	.36	36.13	250.00	-0.00	-0.00
4	6		1	47	.82	49.77	250.00	-0.00	-0.00
5	7		1	45	.94	49.28	250.00	-0.00	-0.00
6	9		1			47.63			
7	8		1			69.83			
8	9		1	37	.83	31.76	150.00	-0.00	-0.00
Gen	Status	Fuel	Pg			Pmin	Pmax	Qmin	Qmax
1	1	COAL	78.13		5.42	10.00	350.00	-300.00	300.00
2	1	COAL	114.18		-9.55	10.00	300.00	-300.00	300.00
3	1	WIND	75.00	_	17.00	0.00	75.00	-300.00	300.00

#### Example using the $\it EMPAR$ solver with multicontingency enabled:

[ExaGO INFO]: Finalizing sopflow application.

\$ ./bin/sopflow -sopflow\_solver EMPAR -sopflow\_enable\_multicontingency 1
[ExaGO INFO]: -- Checking ... -options\_file not passed exists: no
SOPFLOW: Application created
Rank[0]: color = 0, ns = 2, sstart= 0, send=2
SOPFLOW running with 2 scenarios (base case + 1 scenarios)
Rank 0 scenario range [0 -- 2]
SOPFLOW: Using EMPAR solver

```
SCOPFLOW: Application created
SCOPFLOW running with 2 contingencies (base case + 1 contingencies)
Rank 0 has 2 contingencies, range [0 -- 2]
SCOPFLOW: Using IPOPT solver
SCOPFLOW: Setup completed
SCOPFLOW: Application created
SCOPFLOW running with 2 contingencies (base case + 1 contingencies)
Rank 0 has 2 contingencies, range [0 -- 2]
SCOPFLOW: Using IPOPT solver
SCOPFLOW: Setup completed
SOPFLOW: Setup completed
******************
This program contains Ipopt, a library for large-scale nonlinear
optimization. Ipopt is released as open source code under the
Eclipse Public License (EPL).
******************
This is Ipopt version 3.12.10, running with linear solver ma27.
Number of nonzeros in equality constraint Jacobian...:
Number of nonzeros in inequality constraint Jacobian.:
                                                        168
Number of nonzeros in Lagrangian Hessian....:
                                                        215
Total number of variables....:
                                                         51
                   variables with only lower bounds:
                                                          0
               variables with lower and upper bounds:
                                                         35
                   variables with only upper bounds:
                                                         0
Total number of equality constraints....:
                                                         42
Total number of inequality constraints....:
                                                         58
       inequality constraints with only lower bounds:
                                                         2.1
   inequality constraints with lower and upper bounds:
                                                         37
       inequality constraints with only upper bounds:
                                                          \cap
iter objective inf_pr inf_du lq(mu) ||d|| lq(rq) alpha_du alpha_pr ls
0 \quad 7.4421250e + 03 \quad 1.80e + 00 \quad 1.04e + 01 \quad -1.0 \quad 0.00e + 00 \quad -0.00e + 00 \quad 0.00e + 00
1 6.5388630e+03 1.51e+00 2.24e+01 -1.0 1.07e+00 - 5.38e-01 1.58e-01f 1
2 5.9394843e+03 1.28e+00 2.56e+02 -1.0 7.32e-01 - 3.31e-02 1.56e-01f 1
29 3.0556401e+03 1.07e-14 5.62e-10 -8.6 4.33e-09 - 1.00e+00 1.00e+00f 1
Number of Iterations...: 29
                                 (scaled)
                                                         (unscaled)
Objective.....: 6.8512110964678271e+01 3.0556401490246512e+03
                                             2.5079333284691651e-08
Dual infeasibility..: 5.6231688979129259e-10
Constraint violation:
                     1.0658141036401503e-14
                                               1.0658141036401503e-14
```

```
Complementarity....: 2.7880113972390153e-09 1.2434530831686009e-07
Overall NLP error...: 2.7880113972390153e-09 1.2434530831686009e-07
Number of objective function evaluations
                                                   = 48
Number of objective gradient evaluations
                                                   = 30
Number of equality constraint evaluations
                                                  = 48
Number of inequality constraint evaluations
                                                   = 48
Number of equality constraint Jacobian evaluations = 30
Number of inequality constraint Jacobian evaluations = 30
Number of Lagrangian Hessian evaluations
                                                   = 29
Total CPU secs in IPOPT (w/o function evaluations)
                                                   =
                                                          0.010
Total CPU secs in NLP function evaluations
                                                          0.006
EXIT: Optimal Solution Found.
This is Ipopt version 3.12.10, running with linear solver ma27.
Number of nonzeros in equality constraint Jacobian...:
                                                          226
Number of nonzeros in inequality constraint Jacobian.:
                                                          168
Number of nonzeros in Lagrangian Hessian....:
                                                          215
Total number of variables....:
                                                           51
                    variables with only lower bounds:
               variables with lower and upper bounds:
                                                           35
                    variables with only upper bounds:
                                                           0
Total number of equality constraints....:
                                                           42
Total number of inequality constraints....:
                                                           58
        inequality constraints with only lower bounds:
                                                           21
   inequality constraints with lower and upper bounds:
                                                           37
       inequality constraints with only upper bounds:
iter objective inf_pr inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr ls
0 \quad 7.4421250e+03 \quad 1.80e+00 \quad 1.04e+01 \quad -1.0 \quad 0.00e+00 \quad -0.00e+00 \quad 0.00e+00 \quad 0
1 6.2852600e+03 1.53e+00 2.73e+01 -1.0 1.34e+00 - 3.21e-01 1.50e-01f 1
28 2.8488309e+03 1.07e-14 3.60e-09 -8.6 1.94e-08 - 1.00e+00 1.00e+00h 1
Number of Iterations...: 28
                               (scaled)
                                                       (unscaled)
Objective..... 6.3875131478884526e+01
                                                2.8488308639582501e+03
Dual infeasibility..: 3.5985088591418374e-09
                                                1.6049349511772596e-07
Constraint violation: 1.0658141036401503e-14
                                               1.0658141036401503e-14
Complementarity....: 4.7446211848960068e-09
                                                2.1161010484636192e-07
Overall NLP error...: 4.7446211848960068e-09 2.1161010484636192e-07
Number of objective function evaluations
                                                   = 43
```

```
Number of objective gradient evaluations = 29
Number of equality constraint evaluations = 43
Number of inequality constraint evaluations = 43
Number of equality constraint Jacobian evaluations = 29
Number of inequality constraint Jacobian evaluations = 29
Number of Lagrangian Hessian evaluations = 28
Total CPU secs in IPOPT (w/o function evaluations) = 0.016
Total CPU secs in NLP function evaluations = 0.009

EXIT: Optimal Solution Found.
```

#### Example using the Ipopt [7] solver with multicontingency and multiperiod enabled:

```
./bin/sopflow -sopflow_solver IPOPT -sopflow_enable_multicontingency 1\
-scopflow_enable_multiperiod 1 -sopflow_Ns -1 -scopflow_Nc -1
[ExaGO INFO]: -- Checking ... -options_file not passed
                                                     exists: no
SOPFLOW: Application created
Rank[0]: color = 0, ns = 4, sstart= 0, send=4
SOPFLOW running with 4 scenarios (base case + 3 scenarios)
Rank 0 scenario range [0 -- 4]
SOPFLOW: Using IPOPT solver
SCOPFLOW: Application created
SCOPFLOW running with 10 contingencies (base case + 9 contingencies)
Rank 0 has 10 contingencies, range [0 -- 10]
SCOPFLOW: Using IPOPT solver
TCOPFLOW: Application created
TCOPFLOW: Duration = 0.166667 hours, timestep = 5.000000 minutes,
         number of time-steps = 3
TCOPFLOW: Using IPOPT solver
TCOPFLOW: Setup completed
TCOPFLOW: Application created
TCOPFLOW: Duration = 0.166667 hours, timestep = 5.000000 minutes,
         number of time-steps = 3
TCOPFLOW: Using IPOPT solver
TCOPFLOW: Setup completed
SCOPFLOW running with 10 contingencies (base case + 9 contingencies)
Rank 0 has 10 contingencies, range [0 -- 10]
SCOPFLOW: Using IPOPT solver
TCOPFLOW: Setup completed
SCOPFLOW: Setup completed
SOPFLOW: Setup completed
******************
This program contains Ipopt, a library for large-scale nonlinear
optimization. Ipopt is released as open source code under the
Eclipse Public License (EPL).
```

```
*****************
This is Ipopt version 3.12.10, running with linear solver ma27.
Number of nonzeros in equality constraint Jacobian...:
                                                         12678
Number of nonzeros in inequality constraint Jacobian.:
                                                          9636
Number of nonzeros in Lagrangian Hessian....:
                                                         10404
Total number of variables....:
                                                          2688
                    variables with only lower bounds:
                                                            0
               variables with lower and upper bounds:
                                                          1728
                    variables with only upper bounds:
                                                            0
Total number of equality constraints....:
                                                          2223
Total number of inequality constraints....:
                                                          2922
       inequality constraints with only lower bounds:
                                                          648
   inequality constraints with lower and upper bounds:
                                                          2274
       inequality constraints with only upper bounds:
                                                            0
iter objective inf_pr inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr ls
0 \quad 8.0374950e + 05 \quad 2.59e + 00 \quad 6.64e + 01 \quad -1.0 \quad 0.00e + 00 \quad -0.00e + 00 \quad 0.00e + 00
1 7.8996010e+05 2.48e+00 6.98e+01 -1.0 1.43e+00 2.0 2.04e-02 3.97e-02f 1
2 7.7076082e+05 2.31e+00 9.50e+01 -1.0 1.16e+00 2.4 2.69e-02 6.75e-02f 1
197 4.2506331e+05 1.42e-14 3.34e-10 -7.0 2.10e-09 - 1.00e+00 1.00e+00h 1
Number of Iterations...: 197
                               (scaled)
                                                        (unscaled)
Objective..... 9.5305673828307717e+03
                                                 4.2506330527425243e+05
Dual infeasibility..: 3.3401254461705077e-10 Constraint violation: 1.4210854715202004e-14
                                                 1.4896959489920465e-08
                                                 1.4210854715202004e-14
Complementarity....: 9.1353878464689626e-08
                                                 4.0743829795251576e-06
Overall NLP error...: 9.1353878464689626e-08
                                                 4.0743829795251576e-06
Number of objective function evaluations
                                                   = 283
Number of objective gradient evaluations
                                                   = 198
Number of equality constraint evaluations
                                                   = 283
Number of inequality constraint evaluations
                                                   = 283
Number of equality constraint Jacobian evaluations = 198
Number of inequality constraint Jacobian evaluations = 198
Number of Lagrangian Hessian evaluations
                                                   = 197
Total CPU secs in IPOPT (w/o function evaluations)
                                                           2.136
Total CPU secs in NLP function evaluations
                                                           4.421
EXIT: Optimal Solution Found.
[ExaGO INFO]: Finalizing sopflow application.
```

# Appendices

# Appendix A

# Symbol reference

Units of measurement are given in Table (A.1), indices and index sets in Table (A.2), subsets in Table (A.3), special set elements in Table (A.4), and real-valued parameters in Table (A.5).

Table A.1: Units of measurement

Symbol	Description
1	dimensionless. Dimensionless real number quantities are indicated by a unit of 1.
USD	US dollar. Cost, penalty, and objective values are expressed in USD.
h	hour. Time is expressed in h.
pu	per unit. Voltage magnitude is expressed in a per unit system under given base values, and the unit is denoted by pu
rad	radian. Voltage angles are expressed in rad.
MW	megawatt. Real power is expressed in MW.
MVar	megavolt-ampere-reactive. Reactive power is expressed in MVar.
MVA	megavolt-ampere. Apparent power is expressed in MVA.
MW at 1 pu	megawatt at unit voltage. Conductance is expressed in MW at 1 pu, meaning the conductance is such as to yield a real power flow equal to the indicated amount when the voltage is equal to 1 pu
MVar at 1 pu	megavolt-ampere-reactive at unit voltage. Susceptance is expressed in MVar at 1 pu, meaning the susceptance is such as to yield a reactive power flow equal to the indicated amount when the voltage is equal to 1 pu

Table A.2: Index sets

Symbol	Description
$a \in A$ $i \in I$	areas buses

Table A.2: Continued

Symbol	Description
$j \in J$	bus-connected grid components, i.e. loads, shunts, generators, stochastic resources, branches
$k \in K$	security contingencies, i.e. NERC $(n-1)$ - or $(n-k)$ -style contingencies, different from the severe event we are modeling
$s \in S$	stochastic scenarios
$t \in T$	time periods

Table A.3: Subsets

Symbol	Description
$J^{\mathrm{gen}} \subset J$	generators
$J^{\mathrm{ld}}\subset J$	loads
$J^{\mathrm{br}}\subset J$	branches, i.e. lines, transformers
$J^{\mathrm{sh}}\subset J$	shunts
$J_i^{\mathrm{o}} \subset J$	branches with origin bus at bus $i$
$J_i^{\mathrm{d}} \subset J$	branches with destination bus at bus $i$

Table A.4: Special set elements

Symbol	Description
$a_i \in A$	area of bus $i$
$i_j^{\mathrm{d}} \in I$	destination bus of branch $j \in J$
$i_j^{\mathrm{o}} \in I$	origin bus of branch $j \in J^{\text{br}}$

Table A.5: Real-valued parameters

Symbol	Description
$b_j^{\max}$	maximum susceptance for shunt $j \in J^{\text{sh}}$ (MVar at 1 pu)

Table A.5: Continued

Symbol	Description
$b_j^{\min}$	minimum susceptance for shunt $j \in J^{\text{sh}}$ (MVar at 1 pu)
$b_j^{ m ch}$	charging susceptance for branch $j \in J^{\text{br}}$ (MVar at 1 pu)
$b_j^{ m s}$	series susceptance for branch $j \in J^{\text{br}}$ (MVar at 1 pu)
$p_j^{ m gset}$	real power set point of generator $j \in J^{\text{gen}}$ (MW)
$p_j^{ m gmax}$	maximum real power output for generator $j \in J^{\text{gen}}$ (MW)
$p_j^{ m gmin}$	minimum real power output for generator $j \in J^{\text{gen}}$ (MW)
$p_j^{ m r}$	maximum ramp rate for generator $j \in J^{\text{gen}}$ (MW/h)
$q_j^{ m gmax}$	maximum reactive power output for generator $j \in J^{\text{gen}}$ (MVar)
$q_j^{ m gmin}$	minimum reactive power output for generator $j \in J^{\text{gen}}$ (MVar)
$v_i^{\max}$	maximum voltage magnitude for bus $i \in I$ (pu)
$v_i^{\min}$	minimum voltage magnitude for bus $i \in I$ (pu)
$\pi_s$	probability of scenario $s$ (1)

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