





MINISTERIO DE UNIVERSIDADES

Tight and Compact Sample Average Approximation for Joint Chance Constrained Optimal Power Flow

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Outline

Introduction to Joint Chance Constrained Optimal Power Flow (JCC-OPF)

Sample Average Approximation MIP reformulation

Tightening and screening

Valid inequalities

Computational Results: OPF

Introduction to Joint Chance

(JCC-OPF)

Constrained Optimal Power Flow

Introduction to JCC-OPF

- OPF under uncertainty: minimize the expected operating cost whilst guaranteeing that the system withstands unforeseen peeks of electrical load due to stochastic demand.
- Chance constraints ensure feasibility of the system with a tolerable probability of constraint violation.
- Cost-effective decisions can be taken by discarding extreme events or unexpected random circumstances.
- General (linear) formulation:

$$\begin{aligned} & \underset{x}{\text{min}} & c^{\top}x \\ & \text{s.t.} & x \in X \\ & & \mathbb{P}\left\{a_{j}(\omega)^{\top}x \leqslant b_{j}(\omega), \ \forall j\right\} \geqslant 1 - \epsilon. \end{aligned}$$

Introduction to JCC-OPF

Assumptions considered:

- Net demand uncertainty: $\tilde{d}_n = d_n \omega_n$, where d_n where d_n is the predicted value and ω_n is the forecast error with a change of sign.
- **Generation**: To cope with the forecast errors $(\omega_n)_{n \in \mathcal{N}}$, generators' power outputs are adjusted according to the following affine control policy:

$$\tilde{p}_g = p_g - \beta_g \Omega, \quad \forall g \in \mathcal{G},$$

where $\Omega:=\sum_{n\in\mathcal{N}}\omega_n$ is the system-wise aggregated forecast error, and p_g and β_g are the power output dispatch and the participation factor of generating unit g.

Power Balance:

$$\begin{split} \sum_{g \in \mathcal{G}} \tilde{p}_g - \sum_{n \in \mathcal{N}} \tilde{d}_n &= \sum_{g \in \mathcal{G}} \left(p_g - \beta_g \Omega \right) - \sum_{n \in \mathcal{N}} \left(d_n - \omega_n \right) = 0. \\ \sum_{g \in \mathcal{G}} p_g - \sum_{n \in \mathcal{N}} d_n &= 0 \text{ and } \sum_{g \in \mathcal{G}} \beta_g = 1. \end{split}$$

DC Power Flow.

Introduction to JCC-OPF

The mathematical formulation is expressed as follows:

$$\begin{split} & \underset{\rho_{g},\beta_{g}}{\min} \quad \mathbb{E}\left[\sum_{g \in \mathcal{G}} C(\rho_{g},\beta_{g})\right] \\ & \text{s.t.} \sum_{g \in \mathcal{G}} \beta_{g} = 1 \\ & \sum_{g \in \mathcal{G}} p_{g} - \sum_{n \in \mathcal{N}} d_{n} = 0 \\ & \mathbb{P}\left(\begin{array}{c} \underline{p}_{g} \leqslant p_{g} - \Omega \beta_{g} \leqslant \overline{p}_{g}, & \forall g \in \mathcal{G} \\ -\overline{f}_{I} \leqslant \sum_{n \in \mathcal{N}} B_{In} \left(\sum_{g \in \mathcal{G}_{n}} (p_{g} - \Omega \beta_{g}) + \omega_{n} - d_{n}\right) \leqslant \overline{f}_{I}, & \forall I \in \mathcal{L} \end{array}\right) \geqslant 1 - \epsilon, \\ & p_{g}, \beta_{g} \geqslant 0, \forall g \in \mathcal{G}. \end{split}$$

Chance-constrained SAA MIP reformulation (with Big-Ms)

- Sample Average Approx.: assume a finite discrete distribution ⇒ MIP reformulation
- $s \in \mathcal{S}$ scenarios (with equal probabilities)
- $y_s \in \{0,1\}$ such that $y_s = 0 \Leftrightarrow$ the scenario s is satisfied

$$\begin{aligned} & \underset{x}{\min} \quad c^{\top}x \\ & \text{s.t.} \quad x \in X \\ & & \mathbb{P}\left\{a_{j}(\omega)^{\top}x \leqslant b_{j}(\omega), \ \forall j\right\} \geqslant 1 - \epsilon. \end{aligned} \qquad \begin{aligned} & \underset{x}{\min} \quad c^{\top}x \\ & \text{s.t.} \quad x \in X \\ & a_{js}^{\top}x \leqslant b_{j}(\omega) + M_{js}y_{s}, \quad \forall j, s \\ & \sum_{s \in S}y_{s} \leqslant p \\ & y_{s} \in \{0, 1\}, \quad \forall s. \end{aligned}$$

JCC-OPF via Sample Average Approximation

The MIP reformulation of JCC-OPF writes as follows:

$$\begin{split} & \underset{\rho_g,\beta_g}{\text{min}} \quad \mathbb{E}\left[\sum_{g \in \mathcal{G}} C(\rho_g,\beta_g)\right] \\ & \text{s.t.} \sum_{g \in \mathcal{G}} \beta_g = 1 \\ & \sum_{g \in \mathcal{G}} p_g - \sum_{n \in \mathcal{N}} d_n = 0 \\ & - y_s M_{gs}^1 + \underline{p}_g \leqslant p_g - \Omega_s \beta_g \leqslant \overline{p}_g + y_s M_{gs}^2, \quad \forall g, s \\ & - y_s M_{ls}^3 - \overline{f}_l \leqslant \sum_{n \in \mathcal{N}} B_{ln} \left(\sum_{g \in \mathcal{G}_n} \left(p_g - \Omega_s \beta_g\right) - d_n + \omega_{ns}\right) \leqslant \overline{f}_l + y_s M_{ls}^4, \quad \forall l, s \\ & p_g, \beta_g \geqslant 0, \forall g \in \mathcal{G}. \\ & \sum_{s \in \mathcal{S}} y_s \leqslant p \\ & y_s \in \{0, 1\}, \quad \forall s. \end{split}$$

Tightening and screening

Tightening and screening

Algorithm 1 Iterative Coefficient Strengthening (κ iterations)

```
Initialization: k \leftarrow 0, M_{js}^0 \leftarrow \infty.

while k < \kappa do
for j \in \mathcal{J} and s \in \mathcal{S} do
if M_{js}^k > 0 then
1) Tightening phase: Solve
M_{js}^{k+1} \leftarrow \arg\max_{x,y}
s.t.
```

$$\begin{aligned} M_{js}^{k+1} \leftarrow \arg\max_{x,y} \quad & a_{js}^{\top} x - b_{js} \\ \text{s.t.} \quad & x \in X \\ & a_{js}^{\top} x - b_{js} \leqslant M_{js}^{k} y_{s}, \quad \forall j, s \\ & \sum_{s \in S} y_{s} \leqslant p \\ & 0 \leqslant y_{s} \leqslant 1, \quad \forall s. \end{aligned}$$

```
end if  \text{if } M_j^{k+1} < 0 \text{ then}    2) \text{ Screening phase: Eliminate constraint } (j,s) \text{ from the model.}  end if  \text{end for}  Set k \leftarrow k+1.  end while
```

Valid inequalities

Valid inequalities I: generators

For a given g (the process is analogous for the upper bound constraint)

- $\bullet \ \ \rho_{\rm g} \Omega_{\rm s} \beta_{\rm g} \geqslant \underline{\rho}_{\rm g} \quad \Rightarrow \quad \rho_{\rm g} \underline{\rho}_{\rm g} \geqslant \Omega_{\rm s} \beta_{\rm g}$
- \bullet Let us consider a set of scenarios $\Omega:=\{2,-4,4,5,0\}$
- Acceptable violation probability is 40%. Then, if we have 5 scenarios, a constraint can be violated at most in 2 scenarios.

$$\begin{split} & \rho_g - \underline{\rho}_g \geqslant 2 \, \beta_g \\ & \rho_g - \underline{\rho}_g \geqslant -4 \, \beta_g \\ & \rho_g - \underline{\rho}_g \geqslant 4 \, \beta_g \\ & \rho_g - \underline{\rho}_g \geqslant 5 \, \beta_g \\ & \rho_g - \underline{\rho}_\sigma \geqslant 0 \, \beta_g \end{split}$$

^{*}It is like ordering 1-dimensional affine functions without intercept.

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- In a descending order, the constraint with the third largest Ω must be satisfied.
- Therefore, the following constraint is a valid inequality.

$$p_{\rm g} - \underline{p}_{\rm g} \geqslant 2\,\beta_{\rm g}$$

^{*}It is like ordering 1-dimensional affine functions without intercept.

For a given I (the process is analogous for the upper bound constraint)

$$\begin{split} & \sum_{n \in \mathcal{N}} B_{ln} \left(\sum_{g \in \mathcal{G}_n} \left(p_g - \Omega_s \beta_g \right) + \omega_{ns} - d_n \right) \geqslant -\overline{f}_I \\ & \sum_{n \in \mathcal{N}} B_{ln} \left(\sum_{g \in \mathcal{G}_n} p_g - d_n \right) + \overline{f}_I \geqslant \Omega_s \sum_{n \in \mathcal{N}} B_{ln} \sum_{g \in \mathcal{G}_n} \beta_g - \sum_{n \in \mathcal{N}} B_{ln} \omega_{ns} \end{split}$$

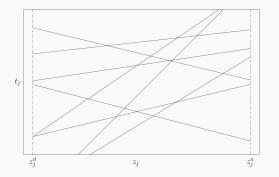
Per each line constraint, the right-hand side is an 1-dimensional affine function per scenario as follows:

$$t_{ls} = \Omega_s z_l + b_{ls}$$
 where:
$$z_l = \sum_{n \in \mathcal{N}} B_{ln} \sum_{g \in \mathcal{G}_n} \beta_g \quad \text{and} \quad \underline{z}_l \leqslant \overline{z}_l$$

$$b_{ls} = \sum_{n \in \mathcal{N}} B_{ln} \omega_{ns}$$

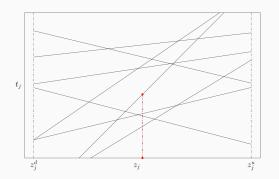
9

$$\sum_{n \in \mathcal{N}} B_{ln} \left(\sum_{g \in \mathcal{G}_n} p_g - d_n \right) + \overline{f}_l \geqslant t_{ls} = \Omega_s z_l + b_{ls}$$

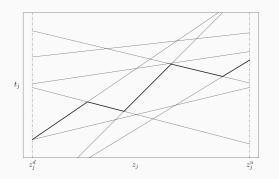


Let us consider an example of 8 scenarios where the constraints can be violated in at most 4 scenarios.

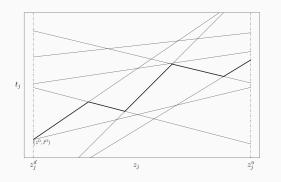
$$\sum_{n \in \mathcal{N}} B_{ln} \left(\sum_{g \in \mathcal{G}_n} p_g - d_n \right) + \overline{f}_l \geqslant t_{ls} = \Omega_s z_l + b_{ls}$$



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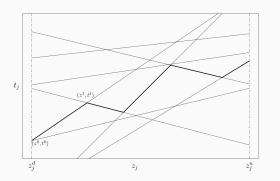


$$\sum_{n \in \mathcal{N}} B_{ln} \left(\sum_{g \in \mathcal{G}_n} p_g - d_n \right) + \overline{f}_l \geqslant t_{ls} = \Omega_s z_l + b_{ls}$$

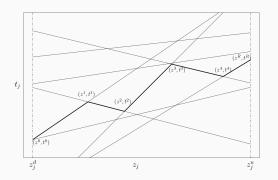


Rider Algorithm

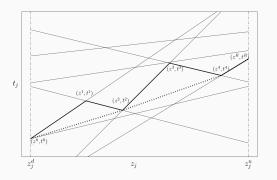
$$\sum_{n \in \mathcal{N}} B_{ln} \left(\sum_{g \in \mathcal{G}_n} p_g - d_n \right) + \overline{f}_l \geqslant t_{ls} = \Omega_s z_l + b_{ls}$$



$$\sum_{n \in \mathcal{N}} B_{ln} \left(\sum_{g \in \mathcal{G}_n} p_g - d_n \right) + \overline{f}_l \geqslant t_{ls} = \Omega_s z_l + b_{ls}$$



$$\sum_{n \in \mathcal{N}} B_{ln} \left(\sum_{g \in \mathcal{G}_n} p_g - d_n \right) + \overline{f}_l \geqslant t_{ls} = \Omega_s z_l + b_{ls}$$



Lower hull: Jarvis March, Graham scan

Computational Results: OPF

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- Approaches **T**, **TS**, **V** and **TS+V** using five standard power systems.
- Instance: IEEE-118 test system: 118 nodes, 19 generators, 186 lines.
- GUROBI 9.1.2 on a Linux-based server with CPUs clocking at 2.6 GHz, 6 threads and 32 GB of RAM.
- 1000 scenarios, 5% violation of the JCC ($\epsilon = 0.05, p = 50$).
- Time limit: 10 hours.
- Results averaged over ten instances.

Computational Results

IEEE-118	BN	T (3)	TS (3)	BN+V	TS (1)+ V
#CON	410413	100%	5.6%	101.6%	2.68%
LRgap	0.956%	0.434%	0.434%	0.4784%	0.2821%
MIPgap	0.29% (0)	0.12% (0)	0.01% (6)	0.03% (1)	0.00% (10)
Time	36000	1.0×	1.4×	1.1×	23.1x

Comparison

	Methods	IEEE-118
	TS+V	0.00%
Average cost increase	CVaR	0.57%
Average cost increase	ALSO-X	0.08%
	ALSO-X+	0.05%
	TS+V	23.1x
Cuandum factor	TS+V CVaR ALSO-X ALSO-X+	4045.5x
Speedup factor	ALSO-X 148.6x	
	$ALSO ext{-}X ext{+}$	49.3x

The End

Á. Porras, C. Domínguez, J.M. Morales, and S. Pineda. (2022) Tight and compact sample average approximation for joint chance-constrained optimal power flow. *arXiv* preprint *arXiv*:2205.03370.



THANK YOU FOR YOUR ATTENTION