Semi-definite relaxations for Optimal Power Flow

A. Isakov, A. Varets, S. Kasymaliev, M. Vinogradov

Skolkovo Institute of Science and Technology

March 22, 2019



Abstract

Problem. Given a task of finding an optimal solution for the Optimal Power Flow (OPF) problem. We have methods to solve it (AC-OPF, DC-OPF), which give local optima and depend on a starter point. Our team wants to try and reformulate this problem as convex one using SDP-relaxation.

Results. We managed to:

- ✓ Learn SDP with solving the problem at 3-bus case
- ✓ Compare result with Newton-Rhapson methods



OPF

Optimal power flow is an optimization problem widely employed in analysis and operation of power systems. It is based on power flow equations: the relationship between voltage input and power injections at buses in an electric power system; the OPF problem itself is stated as minimization of operating cost for an electric power system with network-related and engineering constraints.



classical OPF problem formulation

$$\min_{P_G,Q_G,S,V_d,V_q}\sum_{k\in\mathbb{G}}f_k(P_{Gk})$$

subject to:

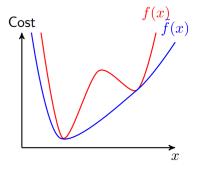
$$\begin{array}{l} \mathsf{P}_{G_{k}}^{min} \leq P_{G_{k}} \leq P_{G_{k}}^{max}, \forall k \in \mathbb{G} \\ \mathsf{Q}_{G_{k}}^{min} \leq Q_{G_{k}} \leq Q_{G_{k}}^{max}, \forall k \in \mathbb{G} \\ (\mathsf{V}_{k}^{min})^{2} \leq \mathsf{V}_{d_{k}}^{2} + \mathsf{V}_{q_{k}}^{2} \leq (\mathsf{V}_{k}^{max})^{2}, \forall k \in \mathbb{N} \\ \|S_{lm} \leq S_{lm}^{max}\|, \forall (l,m) \in \mathbb{L} \\ \mathsf{P}_{Gk} - P_{Dk} = \\ \mathsf{V}_{dk} \sum_{i=1}^{n} (G_{ki} \mathsf{V}_{di} - B_{ki} \mathsf{V}_{qi}) + \mathsf{V}_{qk} \sum_{i=1}^{n} (B_{ki} \mathsf{V}_{di} + G_{ki} \mathsf{V}_{qi}) \\ \mathsf{Q}_{Gk} - Q_{Dk} = \\ \mathsf{V}_{dk} \sum_{i=1}^{n} (-B_{ki} \mathsf{V}_{di} - G_{ki} \mathsf{V}_{qi}) + \mathsf{V}_{qk} \sum_{i=1}^{n} (G_{ki} \mathsf{V}_{di} - B_{ki} \mathsf{V}_{qi}) \\ \forall k \in \mathbb{N} \end{array}$$



SDP

Convex relaxation transforms OPF to convex Semi-Definite Program (SDP).

Under certain conditions, the obtained solution is the global optimum to the original OPF problem.



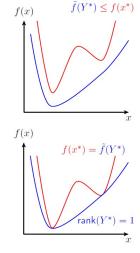


SDP

EXACT: $W = VV^T$ RELAX: W0For the objective functions EXACT_RELAX

If W^* happens with rank(W) = 1 then EXACT

= RELAX





OPF problem formulation in SDP

$$\min_{W} \sum_{k \in \mathbb{G}} \alpha_{k}$$

$$\alpha_{k} = c_{k,2}f^{2}(W) + c_{k,1}f(W) + c_{k,0}, \ f(W) = tr(Z_{k}W) - P_{Dk}$$
subject to:
$$P_{G_{k}}^{min} - P_{Dk} \leq tr(Z_{k}W) \leq P_{G_{k}}^{max} - P_{Dk}, \forall k \in \mathbb{G}$$

$$Q_{G_{k}}^{min} - Q_{Dk} \leq tr(\overline{Z_{k}}W) \leq Q_{G_{k}}^{max} - Q_{Dk}, \forall k \in \mathbb{G}$$

$$(V_{k}^{min})^{2} \leq tr(M_{k}W) \leq (V_{k}^{max})^{2}, \forall k \in \mathbb{N}$$

$$\begin{bmatrix} -(S_{lm}^{max})^{2} & tr(Z_{lm}W) & tr(\overline{Z_{lm}}W) \\ tr(Z_{lm}W) & -1 & 0 \\ tr(\overline{Z_{lm}})W & 0 & -1 \end{bmatrix} \succeq 0, \forall (I, m) \in \mathbb{L}$$

$$W \succ 0$$



Expectations and delivery

☑ Reformulate OPF problem as convex using SDP

 ${f {\it C}}$ Implement SDP for OPF and apply it to artificial case



Results: OPF-SDP solved

We compared our solution with another result obtained via the use of the Newton-Raphson method, namely, ± 5707.11 .

Status: Solved

Optimal value (cvx_optval): +5.80543



Challenges

These are the challenges we've faced during this work:

rhs: -95.0 vexity: Convex.AffineVexity() current status: not yet solved WARNING: Problem status Infeasible; solution may be inaccurate.

6x6 Array{Float64,2}:

NaN MaN NaN NaN



Thank you for your attention!



References

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