# Scaling Security Constrained Optimal Power Flow to Multi-Timestep

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#### **UT Austin**





#### **CU Boulder**



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Event 3: 2nd overall Final Event Outside of prizes

Challenge 3

Event 2 1st overall Event 3 3rd overall Final Event Outside of prizes

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### **Challenge 2**

- 5 minute time-limit
- Up to six nodes available
- Single-time period

### Challenge 3

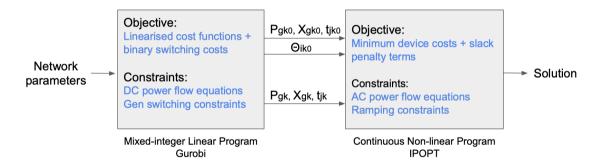
- 10 minute time-limit
- Single node no parallel computing
- Multi-time period up to 48 steps

### Minimum number of decision variables to guarantee feasibility

(no. of devices  $\times$  2 + no. of buses  $\times$  2)  $\times$  no. of timesteps

#### Minimum number of constraints (not including bounds)

(no. of devices + no. of branches  $\times$ 2 + no. of buses  $\times$ 2)  $\times$  no. of timesteps



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## Ramping limits

- 2 Unit commitment decisions up and down time
- 3 Energy constraints

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### Ramping limits

$$egin{aligned} m{
ho}_{j,t} &\leq m{
ho}_{j,t-1} + \Delta \ m{
ho}_{j,t} &\leq m{
ho}_{j,t}^{max} \end{aligned}$$

2 Unit commitment decisions – up and down time

3 Energy constraints

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Keep a set of variables for every time-step

Pros: Maintains whole feasible space

**Cons:** For our approach, even after throwing out unit commitment, this would not reliably finish in 10m for networks larger than 4k.

Finding a single solution that will be feasible for all time-steps

$$p_{j,t}^{min} = \max(p_{j,t}^{min} orall t)$$
  
 $p_{j,t}^{max} = \min(p_{j,t}^{max} orall t)$ 

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Finding a single solution that will be feasible for all time-steps

$$egin{aligned} p_{j,t}^{min} &= \max(p_{j,t}^{min} orall t) \ p_{j,t}^{max} &= \min(p_{j,t}^{max} orall t) \end{aligned}$$

**Pros:** Extremely fast and relatively successful.

**Cons:** Economically inefficient. In rare cases there were instances where  $p_{j,t}^{min} > p_{j,t}^{max}$ .

## Approach 3: Reduced number of time-steps

Break the control horizon into a smaller number of steps

- Pick a time-step mapping (e.g. break the 48 time-steps into four groups)
- Create a set of decision variables for the reduced time-steps

 $egin{aligned} p_{j, au}^{min} &= \max(p_{j,t}^{min} orall t \in au) \quad ext{etc...} \ p_{j, au} &\leq p_{j, au-1} + \Delta \quad ext{etc...} \end{aligned}$ 

**Pros:** Economic improvements over single time-step.

**Cons:** Due to system variables  $v, \theta$  it is not possible to cluster time-steps for each device. Therefore, it is hard to pick the best time-step mapping. Methods for finding the optimal time-steps would eat into available time.

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# Bonus: Speculation about where out mistakes might have been

- (1) Ignoring reserve products initially (and later only adding them in a heuristic way)
- 2 Two-stage linearized approach was less resilient without parallel computation
- 3 Line switching (?)
- 4 Throwing too much out energy constraints, contingencies