

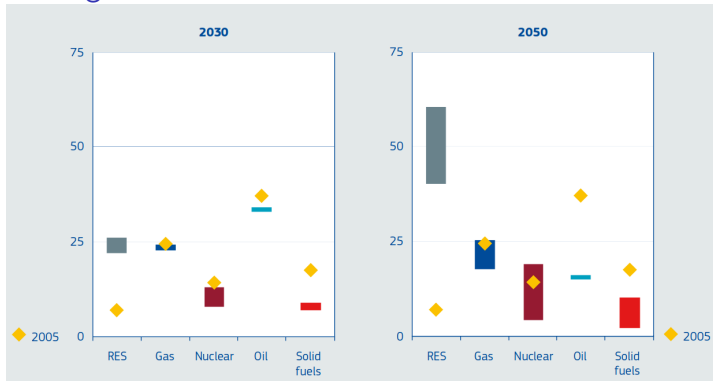
Global Solution of Economic Dispatch with Valve Point Effects and Transmission Constraints

Loïc Van Hoorebeeck
Anthony Papavasiliou
Pierre-Antoine Absil

PSCC 2020

On the place of gas energy in tomorrow's power mix

European targets for 2030 and 2050



“Natural gas will continue to play a key role in the EU’s energy mix in the coming years and gas can gain importance as the *back-up fuel* for variable electricity generation.” (European Commission’s Communication Energy 2020)

Source: European Energy roadmap 2050

1. **Problem statement** - Economic Dispatch with Valve Point Effect.
2. **Methods** - Adaptive piecewise linearization.
 - Exact method
 - Heuristic
3. **Convergence Analysis** - Optimality gap and other bounds.
4. **Study Cases** - modified IEEE-57 and IEEE-118 bus case.
5. **Conclusion**

Problem Statement

2. Problem statement - Economic Dispatch with Valve Point Effect.

Data

Set of bus nodes: N

Set of lines: K

Load demand: D_{nt} $n \in N, t \in T$

Set of producers with cost function f_g $g \in G$

Ramp constraint: R_g^+, R_g^- $g \in G$

Max flow constraint: TC_k $k \in K$

Decision variables

Production: p_{gt} $g \in G, t \in T$

Flow: e_{kt} $k \in K, t \in T$

Problem

How to optimally *dispatch* the power between producers ?

Optimization model

$$\min_{\mathbf{p}, \mathbf{e}} \sum_{g=1, t=1} f_g(p_{gt})$$

Fuel cost minimization

$$\text{s.t. } p_{gt} \leq P_g^+,$$

$$P_g^- \leq p_{gt},$$

$$-\sum_{g \in G_n} p_{gt} - \sum_{k=(\cdot, n)} e_{kt} + D_{nt} + \sum_{k=(n, \cdot)} e_{kt} = 0,$$

$$-R_g^- \leq p_{gt} - p_{g(t-1)} \leq R_g^+,$$

$$-TC_k \leq e_{kt} \leq TC_k.$$

Nomenclature

Bus nodes: N

Cost: f_g

Flow: e_{kt}

Lines: K

Load: D_{nt}

Max flow: TC_k

Production: p_{gt}

Ramp: R_g^+, R_g^-

Optimization model

$$\min_{\mathbf{p}, \mathbf{e}} \sum_{g=1, t=1} f_g(p_{gt})$$

s.t. $p_{gt} \leq P_g^+$, **Restricted power range**

$P_g^- \leq p_{gt}$,

$$- \sum_{g \in G_n} p_{gt} - \sum_{k=(\cdot, n)} e_{kt} + D_{nt} + \sum_{k=(n, \cdot)} e_{kt} = 0 ,$$

$$- R_g^- \leq p_{gt} - p_{g(t-1)} \leq R_g^+ ,$$

$$- TC_k \leq e_{kt} \leq TC_k .$$

Nomenclature

Bus nodes: N

Cost: f_g

Flow: e_{kt}

Lines: K

Load: D_{nt}

Max flow: TC_k

Production: p_{gt}

Ramp: R_g^+ , R_g^-

Optimization model

$$\min_{\mathbf{p}, \mathbf{e}} \sum_{g=1, t=1} f_g(p_{gt})$$

$$\text{s.t. } p_{gt} \leq P_g^+,$$

$$P_g^- \leq p_{gt},$$

Flow conservation

$$-\sum_{g \in G_n} p_{gt} - \sum_{k=(\cdot, n)} e_{kt} + D_{nt} + \sum_{k=(n, \cdot)} e_{kt} = 0,$$

$$-R_g^- \leq p_{gt} - p_{g(t-1)} \leq R_g^+,$$

$$-TC_k \leq e_{kt} \leq TC_k.$$

Nomenclature

Bus nodes: N

Cost: f_g

Flow: e_{kt}

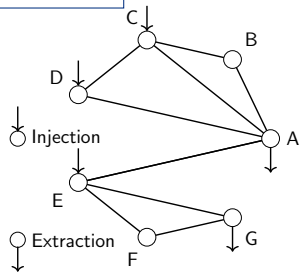
Lines: K

Load: D_{nt}

Max flow: TC_k

Production: p_{gt}

Ramp: R_g^+, R_g^-



Optimization model

$$\min_{\mathbf{p}, \mathbf{e}} \sum_{g=1, t=1} f_g(p_{gt})$$

$$\text{s.t. } p_{gt} \leq P_g^+,$$

$$P_g^- \leq p_{gt},$$

$$-\sum_{g \in G_n} p_{gt} - \sum_{k=(\cdot, n)} e_{kt} + D_{nt} + \sum_{k=(n, \cdot)} e_{kt} = 0,$$

$$-R_g^- \leq p_{gt} - p_{g(t-1)} \leq R_g^+,$$

**Ramp
constraints**

$$-TC_k \leq e_{kt} \leq TC_k.$$

Nomenclature

Bus nodes: N

Cost: f_g

Flow: e_{kt}

Lines: K

Load: D_{nt}

Max flow: TC_k

Production: p_{gt}

Ramp: R_g^+, R_g^-

Optimization model

$$\min_{\mathbf{p}, \mathbf{e}} \sum_{g=1, t=1} f_g(p_{gt})$$

$$\text{s.t. } p_{gt} \leq P_g^+,$$

$$P_g^- \leq p_{gt},$$

$$-\sum_{g \in G_n} p_{gt} - \sum_{k=(\cdot, n)} e_{kt} + D_{nt} + \sum_{k=(n, \cdot)} e_{kt} = 0,$$

$$-R_g^- \leq p_{gt} - p_{g(t-1)} \leq R_g^+,$$

$$-TC_k \leq e_{kt} \leq TC_k.$$

Flow can not exceed the limit

Nomenclature

Bus nodes: N

Cost: f_g

Flow: e_{kt}

Lines: K

Load: D_{nt}

Max flow: TC_k

Production: p_{gt}

Ramp: R_g^+, R_g^-

Optimization model

$$\min_{\mathbf{p}, \mathbf{e}} \sum_{g=1, t=1} f_g(p_{gt})$$

$$\text{s.t. } p_{gt} \leq P_g^+,$$

$$P_g^- \leq p_{gt},$$

$$-\sum_{g \in G_n} p_{gt} - \sum_{k=(\cdot, n)} e_{kt} + D_{nt} + \sum_{k=(n, \cdot)} e_{kt} = 0,$$

$$-R_g^- \leq p_{gt} - p_{g(t-1)} \leq R_g^+,$$

$$-TC_k \leq e_{kt} \leq TC_k.$$

Nomenclature

Bus nodes: N

Cost: f_g

Flow: e_{kt}

Lines: K

Load: D_{nt}

Max flow: TC_k

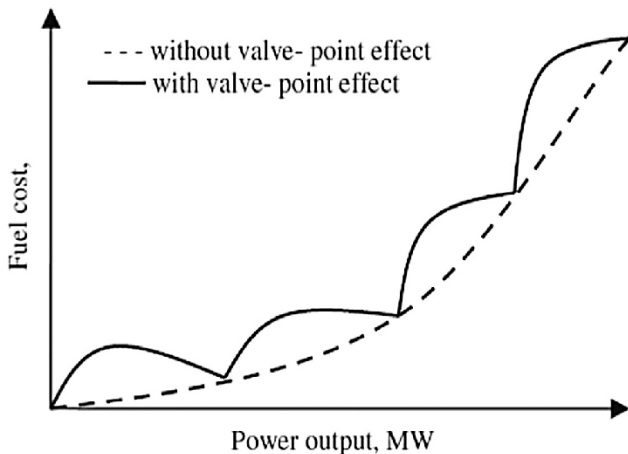
Production: p_{gt}

Ramp: R_g^+, R_g^-

Valve-Point Effect

The VPE is a natural characteristic of a gas turbine. Operating off a valve point increases the throttling losses, therefore rising the heat rate.

$$f_g(p_{gt}) = \underbrace{A_g p_{gt}^2 + B_g p_{gt} + C_g}_{:=f_g^Q(p_{gt})} + \underbrace{D_g |\sin E_g(p_{gt} - P_g^-)|}_{:=f_g^{VPE}(p_{gt})} .$$



Methods

Exact Method

3. Methods - Adaptive piecewise linearization

Original problem

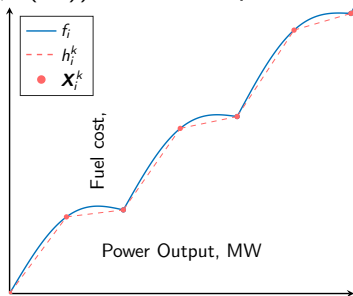
$$(P) \quad \min_{\mathbf{x} \in \Omega} f(\mathbf{x}) = \sum_i f_i(x_i)$$

- ▶ Feasible set is a polyhedron
- ▶ Objective function is non-convex and non-smooth
- ▶ Optimal solution $(\mathbf{x}^*, f(\mathbf{x}^*))$

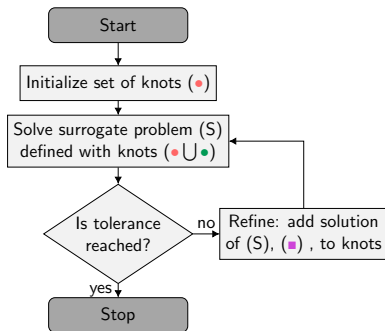
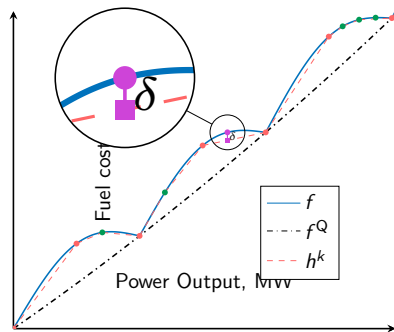
Surrogate problem

$$(S)^k \quad \min_{\mathbf{x} \in \Omega} h^k(\mathbf{x}) = \sum_i h_i^k(x_i)$$

- ▶ Feasible set is unchanged
- ▶ Objective function is piecewise linear defined with *knots* \mathbf{X}^k
- ▶ Optimal solution $(\mathbf{x}_k^{**}, h^k(\mathbf{x}_k^{**}))$



An algorithm for the global solution

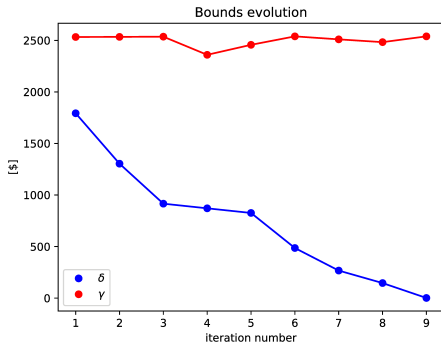


Bounds evolution

Theorem

For Lipschitz continuous cost function f , the sequence of iterations provided by APLA satisfies

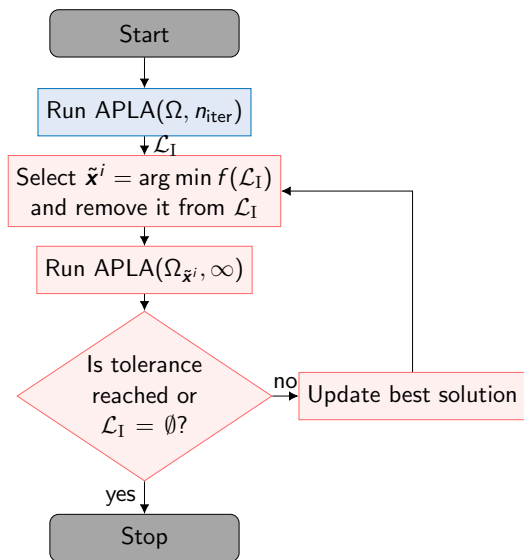
$$\lim_{k \rightarrow \infty} \delta^k = 0.$$



Methods

Heuristic

Heuristic flowchart



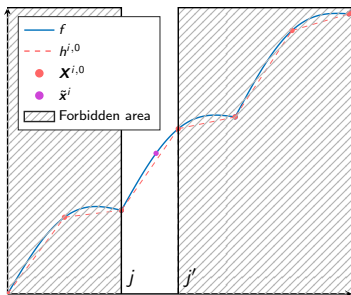
- ▶ Optimization run on the whole domain Ω to improve $\gamma^k, k \in 1 \dots n_{\text{iter}}$.
- ▶ Local improvement of each candidate $\tilde{x}^i \in \mathcal{L}_I$ obtained in previous step. Only δ can be improved.

Domain restriction

Let $\tilde{x}^i = \arg \min f(\mathcal{L}_I)$, $\Omega_{\tilde{x}^i}$ is the local restriction of the domain around \tilde{x}^i . The power generation range becomes

$$X_{gtj}^i \leq p_{gt} \leq X_{gtj'}^i$$

for $j < j'$.

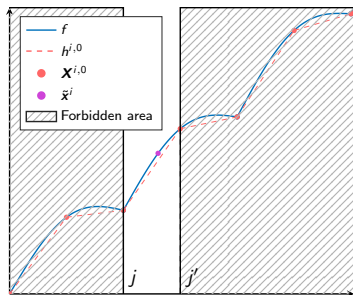


Domain restriction

The surrogate problem $(S)^{i,\tilde{k}}$ becomes

$$\begin{array}{ll} (S)^{i,\tilde{k}} & \min h^{i,\tilde{k}}(\mathbf{x}) \\ \text{subject to} & \mathbf{x} \in \Omega_{\tilde{\mathbf{x}}^i} \end{array}$$

- ▶ **Easier** as the number of integer variables decreases drastically ($j = j' - 1 \Rightarrow (S)^{i,0}$ is a (LP))
- ▶ Global optimal solution may lie **outside** $\Omega_{\tilde{\mathbf{x}}^i}$
- ▶ Any lower bound is **not** a valid lower bound for (P)



Domain restriction

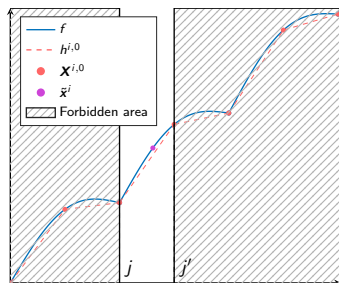
Local heuristic - H-local

"Full" heuristic - H-full

$$(S)^{i,\tilde{k}} \quad \min_{x \in \Omega_{\tilde{x}^i}} h^{i,\tilde{k}}(x)$$

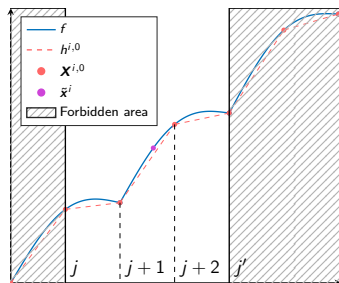
▶ Single intervalle support

▶ $j' = j + 1$.



▶ Multi intervalles support

▶ $j' = j + 3$.



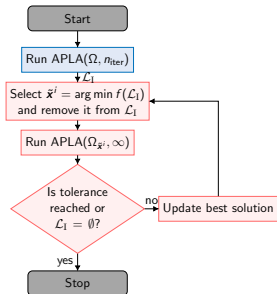
Convergence Analysis

Optimality gap of the heuristic solution (i)

- ▶ Objective value at iteration \tilde{k} of subproblem i

$$f(\mathbf{x}^{i, \tilde{k}}) - f(\mathbf{x}^*) \leq \overbrace{f(\mathbf{x}^{i, \tilde{k}}) - h^{n_{\text{iter}}}(\mathbf{x}^{n_{\text{iter}}})}^{\text{Optimality gap}} + \gamma^{n_{\text{iter}}}$$

- ▶ Surrogate value at iteration n_{iter} on whole domain
- ▶ Solver tolerance

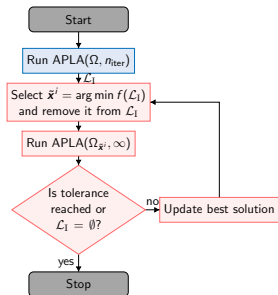


Optimality gap of the heuristic solution (i)

- ▶ Objective value at iteration \tilde{k} of subproblem i

$$f(\mathbf{x}^{i, \tilde{k}}) - f(\mathbf{x}^*) \leq \overbrace{f(\mathbf{x}^{i, \tilde{k}}) - h^{n_{\text{iter}}}(\mathbf{x}^{n_{\text{iter}}})}^{\text{Optimality gap}} + \gamma^{n_{\text{iter}}}$$

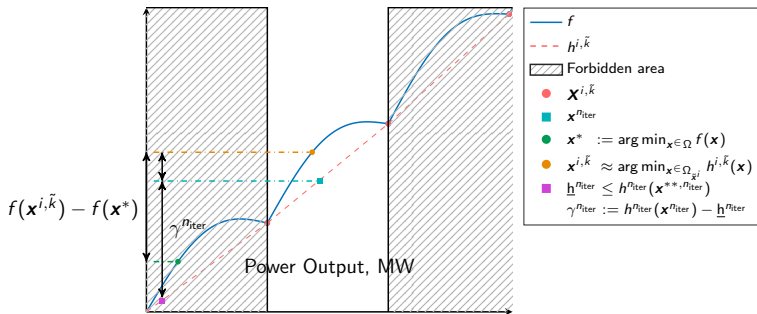
- ▶ Surrogate value at iteration n_{iter} on whole domain
- ▶ Solver tolerance



No convergence guarantee
Once we reach the **inner loop**, the **lower bound** remains unchanged.

A practical visualization (i)

$$f(\mathbf{x}^{i,\tilde{k}}) - f(\mathbf{x}^*) \leq f(\mathbf{x}^{i,\tilde{k}}) - h^{n_{\text{iter}}}(\mathbf{x}^{n_{\text{iter}}}) + \gamma^{n_{\text{iter}}} = f(\mathbf{x}^{i,\tilde{k}}) - \underline{h}^{n_{\text{iter}}}$$

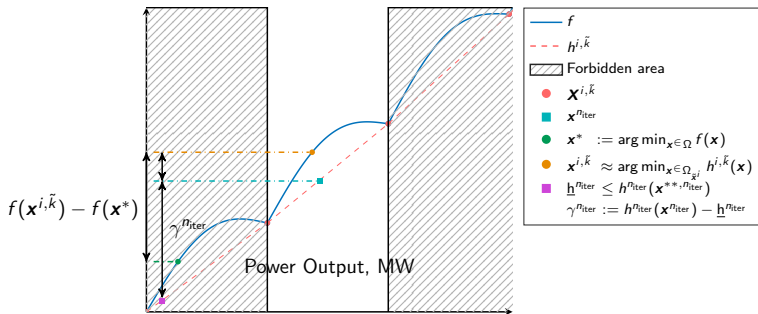


$$\underline{h}^{n_{\text{iter}}} \leq h^{n_{\text{iter}}}(\mathbf{x}^{**}, n_{\text{iter}}) \leq h^{n_{\text{iter}}}(\mathbf{x}^*) \leq f(\mathbf{x}^*)$$

$$-f(\mathbf{x}^*) \leq -\underline{h}^{n_{\text{iter}}}$$

A practical visualization (i)

$$f(\mathbf{x}^{i,\tilde{k}}) - f(\mathbf{x}^*) \leq f(\mathbf{x}^{i,\tilde{k}}) - h^{n_{\text{iter}}}(\mathbf{x}^{n_{\text{iter}}}) + \gamma^{n_{\text{iter}}} = f(\mathbf{x}^{i,\tilde{k}}) - \underline{h}^{n_{\text{iter}}}$$

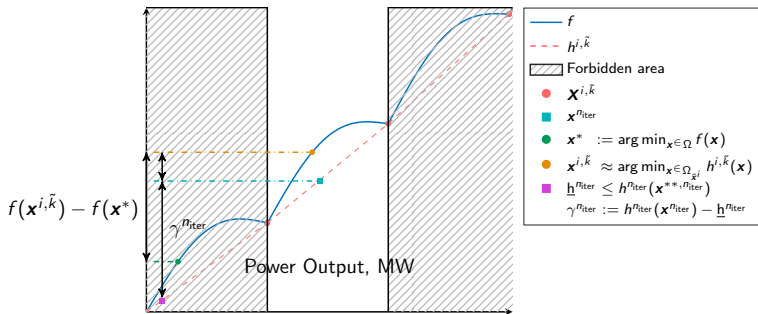


$$\underline{h}^{n_{\text{iter}}} \leq h^{n_{\text{iter}}}(\mathbf{x}^{**}, n_{\text{iter}}) \leq h^{n_{\text{iter}}}(\mathbf{x}^*) \leq f(\mathbf{x}^*)$$

$$-f(\mathbf{x}^*) \leq -\underline{h}^{n_{\text{iter}}}$$

A practical visualization (i)

$$f(\mathbf{x}^{i,\tilde{k}}) - f(\mathbf{x}^*) \leq f(\mathbf{x}^{i,\tilde{k}}) - h^{n_{\text{iter}}}(\mathbf{x}^{n_{\text{iter}}}) + \gamma^{n_{\text{iter}}} = f(\mathbf{x}^{i,\tilde{k}}) - \underline{h}^{n_{\text{iter}}}$$



$$\underline{h}^{n_{\text{iter}}} \leq h^{n_{\text{iter}}}(\mathbf{x}^{*,n_{\text{iter}}}) \leq h^{n_{\text{iter}}}(\mathbf{x}^*) \leq f(\mathbf{x}^*)$$

$$-f(\mathbf{x}^*) \leq -\underline{h}^{n_{\text{iter}}}$$

Optimality gap of the heuristic solution (ii)

Could we obtain an expression in term of δ ?

$$f(\mathbf{x}^{i,\tilde{k}}) - f(\mathbf{x}^*) = f(\mathbf{x}^{i,\tilde{k}}) - f(\mathbf{x}^{*,i}) + \underbrace{f(\mathbf{x}^{*,i}) - f(\mathbf{x}^*)}_{:=\zeta^i}$$

$$\leq \delta^{i,\tilde{k}} + \gamma^{i,\tilde{k}} + \zeta^i \quad (1)$$

$$\approx \delta^{i,\tilde{k}} + \zeta^i \quad (2)$$

Nomenclature

$$\mathbf{x}^* := \arg \min_{\mathbf{x} \in \Omega} f(\mathbf{x})$$

$$\mathbf{x}^{*,i} := \arg \min_{\mathbf{x} \in \Omega_{\tilde{x}^i}} f(\mathbf{x})$$

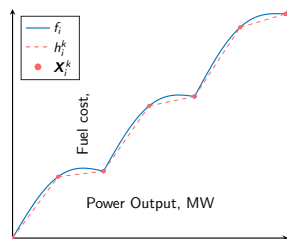
- (1) APLA applied to the restricted domain $\Omega_{\tilde{x}}$ is a valid instance.
- (2) $(S)^{i,\tilde{k}}$ solved to optimality.

Comparison between APLA and the heuristic

APLA

$$(S)^k \quad \min_{\mathbf{x} \in \Omega} h^k(\mathbf{x})$$

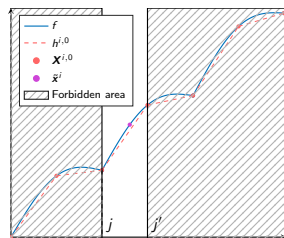
- ▶ $f(\mathbf{x}^k) - f(\mathbf{x}^*) \leq \delta^k + \gamma^k$
- ▶ $\lim_{k \rightarrow \infty} f(\mathbf{x}^k) - f(\mathbf{x}^*) \leq \gamma$
- ▶ Global optimal solution obtained up to the solver accuracy



Heuristic

$$(S)^{i,\tilde{k}} \quad \min_{\mathbf{x} \in \Omega_{\tilde{x}^i}} h^{i,\tilde{k}}(\mathbf{x})$$

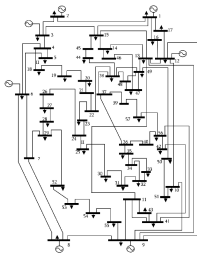
- ▶ $f(\mathbf{x}^{i,\tilde{k}}) - f(\mathbf{x}^*) \leq \delta^{i,\tilde{k}} + \zeta^i$
- ▶ $\lim_{\tilde{k} \rightarrow \infty} f(\mathbf{x}^{i,\tilde{k}}) - f(\mathbf{x}^*) \leq \zeta^i$
- ▶ Best solution bounded by $\zeta^i \geq 0$



Study Cases

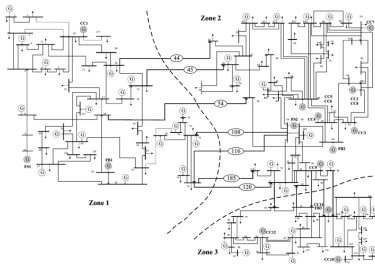
IEEE-57 bus system

- ▶ 57 nodes
- ▶ 7 generators
- ▶ 80 lines
- ▶ 24 time steps



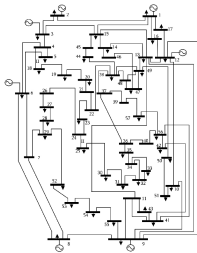
IEEE-118 bus system

- ▶ 118 nodes
- ▶ 54 generators
- ▶ 186 lines
- ▶ 24 time steps



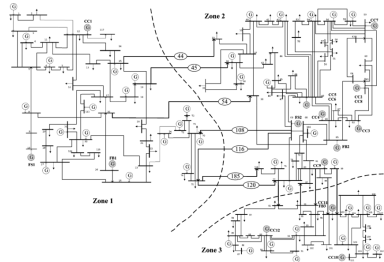
Modified IEEE-57 bus system

- ▶ 57 nodes
- ▶ 7 generators + 10 generators with VPE
- ▶ 80 lines
- ▶ 24 time steps



Modified IEEE-118 bus system

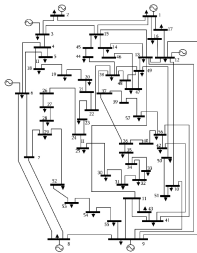
- ▶ 118 nodes
- ▶ 54 generators + 10 generators with VPE
- ▶ 186 lines
- ▶ 24 time steps



? Where to add the VPE generators ?

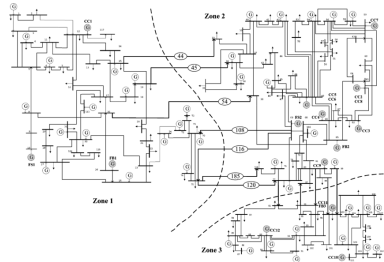
Modified IEEE-57 bus system

- ▶ 57 nodes
- ▶ 7 generators + 10 generators with VPE
- ▶ 80 lines
- ▶ 24 time steps



Modified IEEE-118 bus system

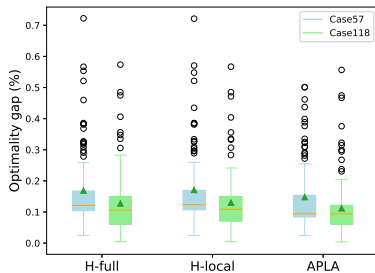
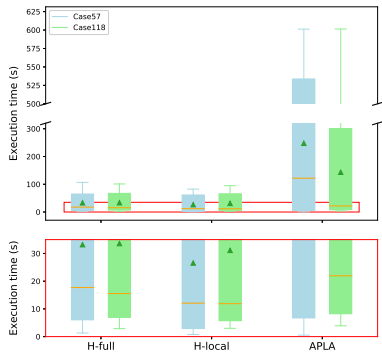
- ▶ 118 nodes
- ▶ 54 generators + 10 generators with VPE
- ▶ 186 lines
- ▶ 24 time steps



? Where to add the VPE generators ?
100 randomly independent trials

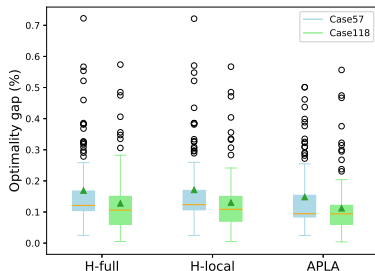
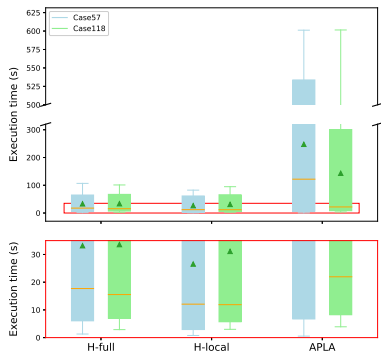
Execution time and optimality gap

100 independent runs with $n_{\text{iter}} = 1$, $\gamma = 0.1\%$ and time limitation of 10 iterations of maximum 60 seconds



Execution time and optimality gap

100 independent runs with $n_{\text{iter}} = 1$, $\gamma = 0.1\%$ and time limitation of 10 iterations of maximum 60 seconds



The heuristics run *significantly* faster for comparable finale optimality gap.

Valve Point Effect impact

QP - Solve (P) while *ignoring* the VPE: a simple Quadratic Programming (convex) problem

Table: Objective mean of each method.

	H-full	H-local	APLA	QP
Case57	621622	621634	621547	654535
Case118	2447475	2447540	2447383	2471132

$$f_g(p_{gt}) = \underbrace{A_g p_{gt}^2 + B_g p_{gt} + C_g}_{:=f_g^Q(p_{gt})} + \underbrace{D_g |\sin E_g(p_{gt} - P_g^-)|}_{:=f_g^{VPE}(p_{gt})}$$

Valve Point Effect impact

QP - Solve (P) while *ignoring* the VPE: a simple **Q**uadratic **P**rogramming (convex) problem

Table: Objective mean of each method.

	H-full	H-local	APLA	QP
Case57	621622	621634	621547	654535
Case118	2447475	2447540	2447383	2471132

Best objectives

Valve Point Effect impact

QP - Solve (P) while *ignoring* the VPE: a simple **Q**uadratic **P**rogramming (convex) problem

Table: Objective mean of each method.

	H-full	H-local	APLA	QP
Case57	621622	621634	621547	654535
Case118	2447475	2447540	2447383	2471132

Nearly as good

⚠ APLA reaches a 10% better optimality gap **but** comparable objective

Valve Point Effect impact

QP - Solve (P) while *ignoring* the VPE: a simple Quadratic Programming (convex) problem

Table: Objective mean of each method.

	H-full	H-local	APLA	QP
Case57	621622	621634	621547	654535
Case118	2447475	2447540	2447383	2471132

Poorer results (5% and 1%)

Conclusion

Conclusion

- ▶ APLA and both heuristics manage to find a **good** candidate ...
... along with a **lower bound**.
- ▶ The heuristics take significantly less time and reach \sim same objective.
- ▶ The VPE *matters* in the studied instances.

Conclusion

- ▶ APLA and both heuristics manage to find a **good** candidate ...
... along with a **lower bound**.
- ▶ The heuristics take significantly less time and reach \sim same objective.
- ▶ The VPE *matters* in the studied instances.

Contact

✉ Loic.vanhoorebeek@uclouvain.be

🌐 <https://perso.uclouvain.be/loic.vanhoorebeek>