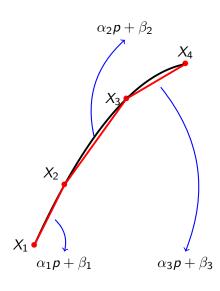
# MILP-Based Algorithm for the Global Solution of Dynamic Economic Dispatch with Valve-Point Effects

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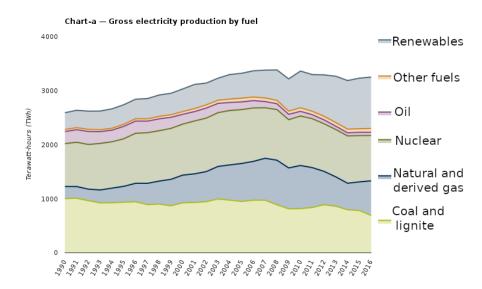






- 1. **Introduction** On the use of gas energy in today's and tomorrow's power mix.
- 2. Problem statement Economic Dispatch with Valve Point Effect.
- 3. **Description of the algorithm** An Adaptive Piecewise-Linear Approximation.
- 4. **Study case** A 10-units dispatch over 24 hours.
- 5. Further work.

# On the place of gas energy in today's power mix



# On the place of gas energy in today's power mix

European plan on climate change consists in the 20-20-20 targets: By 2020...

- ▶ Reduce by 20% the emissions of greenhouse gases (GHB) compared to 1990 levels;
- ► Reach 20% of renewables energy;
- ▶ Increase by 20% the energy efficiency.

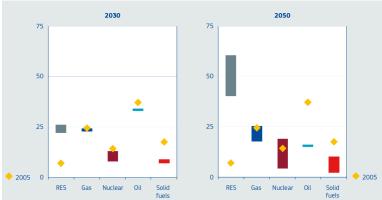
# On the place of gas energy in today's power mix

European plan on climate change consists in the 20-20-20 targets: By 2020...

- √ Reduce by 20% the emissions of greenhouse gases (GHB) compared to 1990 levels; OK since 2016
- ▶ Reach 20% of renewables energy; Still on track: 17.52 % in 2017
- X Increase by 20% the energy efficiency. Not on track: 4 % above objective in 2016

# 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: Energy roadmap 2050

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

#### Data

#### Decision variables

Production:  $p_{it}$   $i \in I, t \in T$ Reserve:  $s_{it}$   $i \in I, t \in T$ 

#### **Problem**

How to optimally dispatch the power between producers ?

#### Valve-Point Effect

Fuel cost,

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

$$f(p) = a p^2 + b p + c + d \left| \sin e(p - P^{\min}) \right|$$
--- without valve- point effect
with valve- point effect

Power output, MW

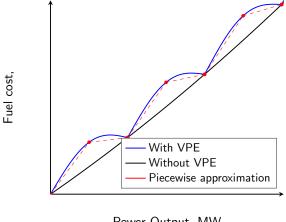
#### Proposed approach

#### Adaptive Piecewise-Linear Under-Approximation

♀ Idea: a sequence of piecewise approximations.

We could use an uniform grid ...

... but there are too many integer variables !



Power Output, MW

# 3. Description of the algorithm - An Adaptive Piecewise-Linear Approximation.

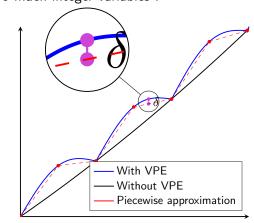
#### Adaptive Piecewise-Linear Under-Approximation

 $\mathbf{\hat{V}}$  Idea: a sequence of piecewise approximations.

We could use an uniform grid ...

Fuel cost

... but there are too much integer variables !



# Proposed approach

# Piecewise linearization of objective

#### First model: binary variables

$$g(\rho, \boldsymbol{\xi}, \boldsymbol{\eta}) := \begin{cases} \sum_{j=1}^{n^{\mathsf{knot}}-1} \alpha_{j} \xi_{j} + \eta_{j} \beta_{j} \,, \\ & \mathsf{with} \ \sum_{j=1}^{n^{\mathsf{knot}}-1} \xi_{j} = \rho \,, \\ & \sum_{j=1}^{n^{\mathsf{knot}}-1} \eta_{j} = 1 \,, \\ & X_{j} \, \eta_{j} \leq \xi_{j} \leq X_{j+1} \, \eta_{j} \,, \\ & \eta_{j} \ \mathsf{binary} \,. \end{cases}$$

 $\alpha_2 p + \beta_2$ 

# Optimality gap

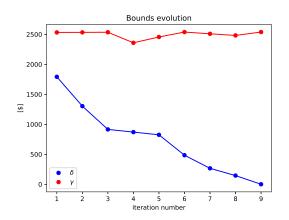
Solver tolerance error Optimality gap  $|f(\mathbf{p}^*) - \mathbf{f}(\mathbf{p}^k)| \leq \delta^k + \gamma^k + \epsilon^k$ 

Over-approximation error

#### What about the convergence ?

- $\gamma^k$  is bounded below by  $\gamma f(\mathbf{p}^*)$ ;
- $\epsilon^k$  is virtually negligible since the "convex zones" are smaller than 0.1% of the domain ;
- $\delta^k$  converges to zero.

#### A practical example



# What about the convergence ?

#### Theorem

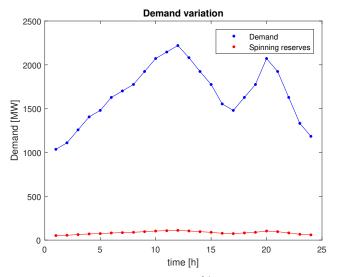
$$\lim_{k\to\infty}\delta^k=0$$

#### Proof (sketch).

- $f_i$  is L-continuous:  $|f_i(p+\Delta) f_i(p)| \le (2a_i P_i^{\mathsf{max}} + b_i + d_i e_i) \Delta$
- ▶  $f = \sum_{i=1}^{n} f_i$  is L-continuous with constant  $K := \sum_{i=1}^{n} K_i$
- ▶ g is also Lipschitz continuous with same constant

$$\begin{split} 0 & \leq \delta^k = \min_{l \in 1 \dots k} f(\mathbf{p}^l) - g^k(\mathbf{p}^k) \\ & \leq f(\mathbf{p}^k) - g^k(\mathbf{p}^k) \\ & = f(\mathbf{p}^k) - g^k(\mathbf{p}^{k-1}) + g^k(\mathbf{p}^{k-1}) - g^k(\mathbf{p}^k) \quad \text{(trick: } +1 -1) \\ & = f(\mathbf{p}^k) - f(\mathbf{p}^{k-1}) + g^k(\mathbf{p}^{k-1}) - g^k(\mathbf{p}^k) \quad \text{(knot updating)} \\ & \leq 2K \left| \left| \mathbf{p}^k - \mathbf{p}^{k-1} \right| \right| \quad \text{(Lipschitz continuity)} \end{split}$$

# 4. Study case - A 10-units dispatch over 24 hours



Spinning reserves set at 5% of the demand. 10 units with valve-point loading effect.

# Results table Previous results

#### And our approach?

TABLE III
SUMMARY RESULTS FOR THE 10-UNIT SYSTEM WITHOUT LOSS

Method		generation c		S-time(min)
Method	Minimum	Average	Maximum	S-unc(min)
SQP [3]	1051163	NA	NA	0.42
EP 3	1048638	NA	NA	15.05
CDE [16]	1036756	1040586	1452558	0.20
GA [12]	1033481	1038014	1042606	3.59
EP-SQP [3]	1031746	1035748	NA	7.26
AIS-SQP 28	1029900	NA	NA	NA
MHEP-SQP[27]	1028924	1031179	NA	21.23
DGPSO [4]	1028835	1030183	NA	4.81
PSO [12]	1027679	1031716	1034340	3.85
SOA 29	1023946	1026289	1029213	NA
IPSO 8	1023807	1026863	NA	0.05
CSDE [17]	1023432	1026475	1027634	0.3
CE [14]	1022702	1024024	NA	0.33
ECE [14]	1022272	1023334	NA	0.33
AIS [13]	1021980	1023156	1024973	25.35
ABC 12	1021576	1022686	1024316	3.47
CDBCO [23]	1021500	1024300	NA	0.73
SOA-SQP[29]	1021460	1023841	1026852	NA
AHDE 10	1020082	1022474	1024484	1.10
CDE 16	1019123	1020870	1023115	0.32
HHS [15]	1019091	NA	NA	10.19
ICPSO 9	1019072	1020027	NA	0.35
CSAPSO[III	1018767	1019874	NA	0.350
EAPSO [18]	1018510	1018701	1019302	0.63
HIGA [30]	1018473	1019328	1022284	4.41
ICA [20]	1018467	1019291	1021796	NA
TVAC-IPSO 19	1018217	1018965	1020418	2.72
HBPSO[31]	1018159	1019850	1021813	3.09
CSO [25]	1017660	1018120	1019286	0.90
EBSO [21]	1017147	1017526	1017891	0.15
MILP	1016316			0.94
MILP-IPM	1016311			1.02

NA denotes that the value was not available in the literature.

Pan, Jian, and Yang, "A hybrid MILP and IPM approach for dynamic economic dispatch with valve-point effects", 2018.

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#### And our approach?

APLUA 1016276\$ 15(min)  APLUA 1016207\$ 1.5(min)  + Local Heuristic			
+ Local	APLUA	1016276\$	15(min)
	+ Local	1016207\$	1.5(min)

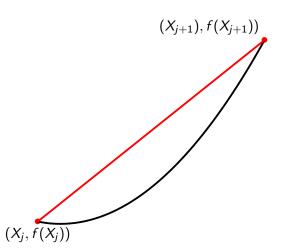
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#### 5. Further work

Important condition of the proof: Under-approximation

 $\Rightarrow$  The proof is not valid for convex functions (e.g. without valve point effect)



Important condition of the proof: **Under**-approximation

- ▶ Possible to prove that  $g^{k+1} \ge g^k$  and that we cannot do better with that number of points
- Number of integer variables rises linearly ( $\sim$  factor 2)

# Further Work: power losses and network constraints

(Revisited) demand constraints

Network constraints

$$\sum_{i=1}^n \rho_{it} = D_t + \rho^L(\mathbf{p}_t)$$

 $p^L(\mathbf{p}_t)$  models the transmission losses computed as

$$\rho^L(\mathbf{p}_t) = \mathbf{p}_t^T \mathbf{B} \mathbf{p}_t + \mathbf{B}_0 \mathbf{p}_t + \mathbf{B}_{00}$$

with **B** symmetric matrix.

#### Conclusion

- ► APLUA manages to find a good candidate ...
- ▶ ... it takes more time but same range number ...
- ... and we are limited by the solver tolerance gap.
- ? How to take the quadratic transmission lost into account?

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#### Contact

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