



Combinatorial optimization methods for energy management

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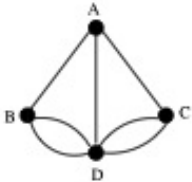
Combinatorial optimization

Operations research (maths + algorithmics + computer sc.) to solve decision pbs

Combinatorial explosion : $O(n!)$ solutions

20 nodes $\approx 1e^{17}$ solutions

> "Brute-Force" Method



Nodes		
10		
15		
19		
27		
35		
40		
50		

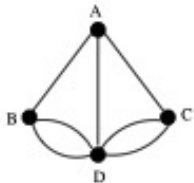
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- > "Brute-Force" Method
- > Proc. 3GHz : 3 op / nano second



Nodes	Proc. 3 GHz	
10	1/100s	
15	1 h	
19	1 an	
27	8× age univers	
35	?	
40	?	
50	?	

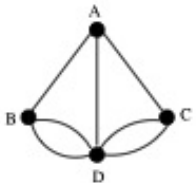
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- > Proc. Planck : 1 op / Planck time (5.39×10^{-44})s



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35	?	5/1000s
40	?	12 ans
50	?	4000× age univers

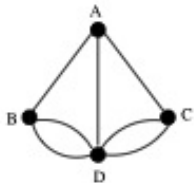
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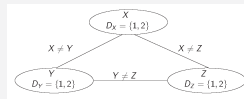
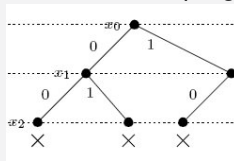
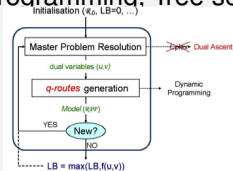
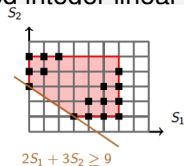
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Theoretical analysis of classes of problems and generic solution methods

Solving combinatorial optimization problems

Powerful tools and methods

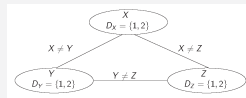
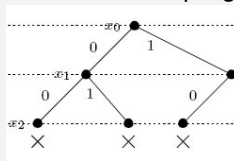
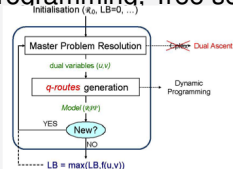
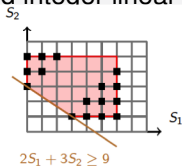
Mixed integer linear programming, Tree search, Constraint programming, etc.



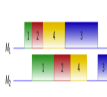
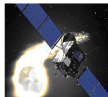
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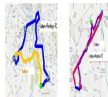
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Various applications



6	1	4	5	
2	8	3	5	6
8	6	4	7	3
7	0	1	3	6
5	7	2	6	9
4	1	5	8	7

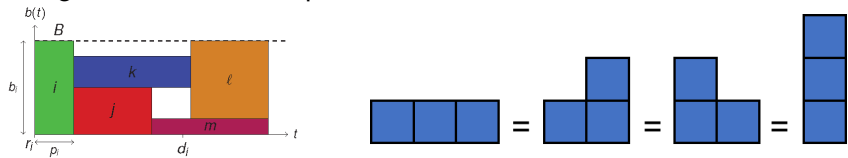


When tackling a practical problem, there are two elements to consider : the **modeling** phase and the **solving** phase (decomposition methods, ...)

LAAS research on energy and combinatorial problems

Before 2008-2010 :

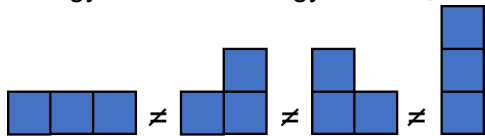
- > Energetic resources : equivalent to man-hours / materials / machines



Energetic reasoning (P. Lopez 90s), RCPSP with energy resources (C. Artigues et al.), etc.

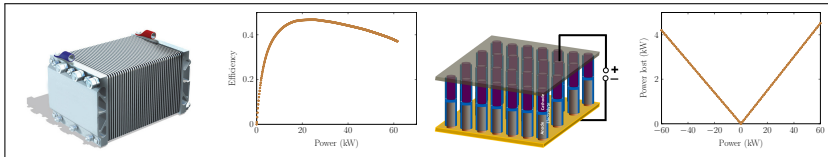
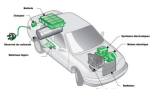
After 2010

- > Energy sources : energy transfer, energy losses, dynamics

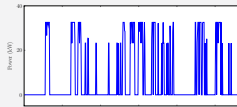


Energy optimization in hybrid electric vehicles

Energy sources characteristics : **power limits** (kW), **efficiency** (%), **capacity** (kWh) ...

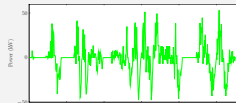


Find per instant the **optimal power split** between energy sources to **minimize** the **total fuel consumption**.



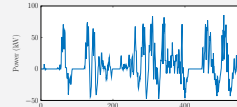
Power from FC

+



Power from/to SE

=

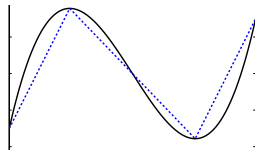


Total power provided

Modeling phase

- > Intrinsic **non linearities** : non linear conversion/consumption functions
 - approximate with piecewise linear functions

- MINLP solution methods

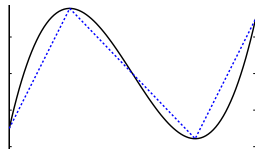


Solving phase

Modeling phase

- > Intrinsic **non linearities** : non linear conversion/consumption functions
 - approximate with piecewise linear functions
- + (more) tractable problems
- try and error approach : No guarantees on the solution quality, undefined number of iterations
- global optimality cannot be guaranteed

- MINLP solution methods



Solving phase

Modeling phase

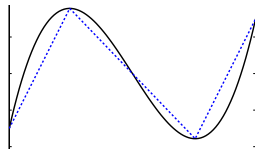
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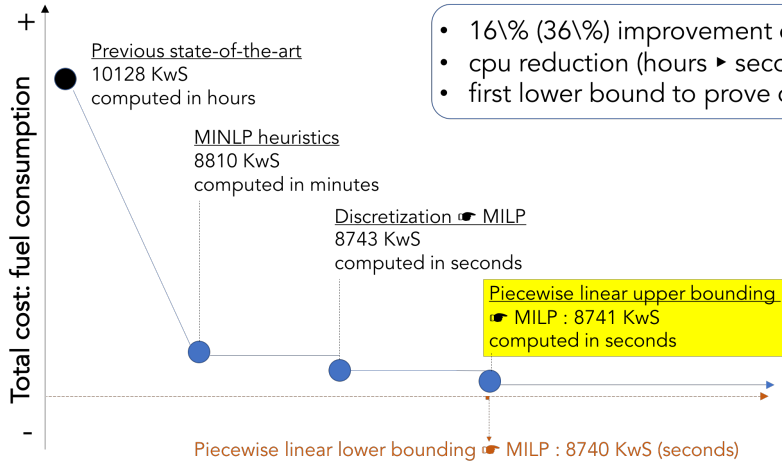
- MINLP solution methods

- + global optimality guaranteed if carried out to completion
- only for small/medium instances



Solving phase

Results obtained for hybrid electric vehicles

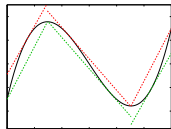
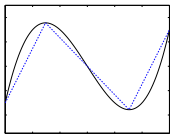


- 16\% (36\%) improvement over state-of-the-art
- cpu reduction (hours ▶ seconds)
- first lower bound to prove optimality gap 0.01\%

How it works, Why it works

New two-step solution scheme : (Ngueveu et al., 2014, 2016, 2018)

Step 1 : Piecewise linear **bounding** of the nonlinear energy transfer/efficiency functions



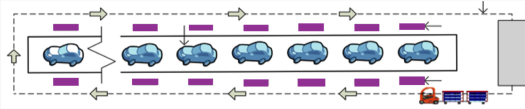
Step 2 : Reformulation of the problem into **two** mixed integer problems (MILP)

- > solve with a black box MILP solver
- > or design a dedicated solution method (**only one needed**)

3rd Robert Faure ROADEF prize 2018 (<https://www.laas.fr/public/fr/node/1741>)

What happens if the problem is not easy to solve ?

Energy aware-feeding system (production line)



minimize Energy consumption and Work-in Process

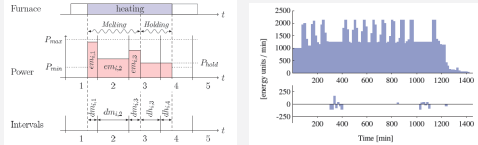
Energetic resources w non linear conversion functions



constraint propagation algorithms : "Energy reasoning"

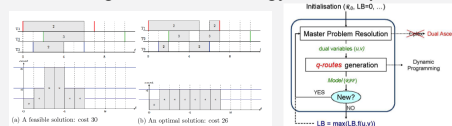
C. Artigues, C. Briand, E. Hébrard, N. Jozefowicz, P. Lopez, S. U. Nguveu + A. Haït (ass. res.).
4 PhD : M. Guemri (2013), Y. Gaoua (2014), M. Nattaf (2016), Y. He (2017). 1 postdoc : G. Simonin (2012–2014) Several interships

Production scheduling with energy costs, case study : foundry (Min energy cost) / steel plant (load tracking)



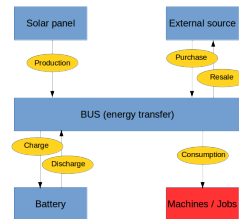
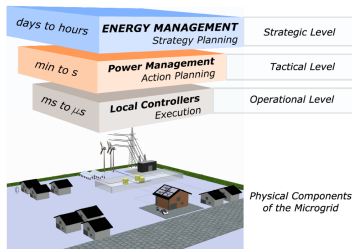
hybrid MILP/CP method, mixed continuous/discrete models

Scheduling at minimal energy consumption cost



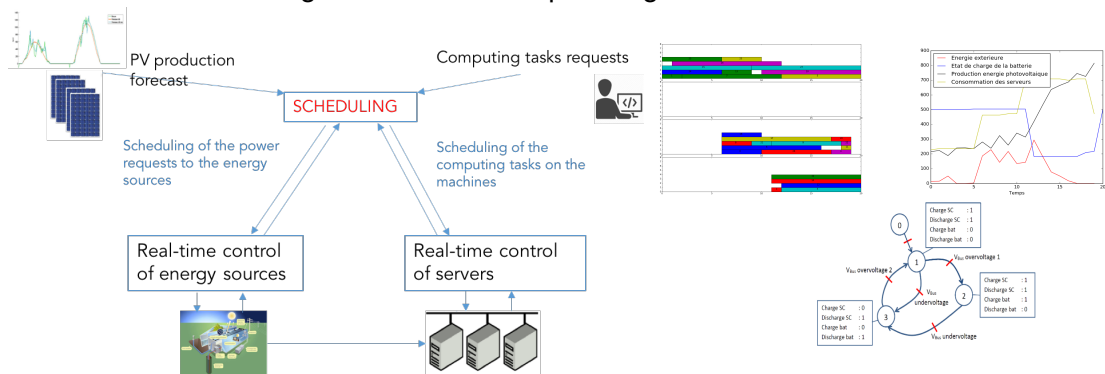
Dantzig-Wolfe decomposition, column generation method

Our interest : Long term / short term planning / link with control

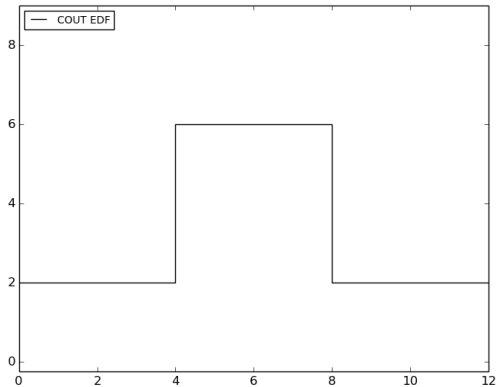


Poster and demonstrations tomorrow during your visit of the platform 😊

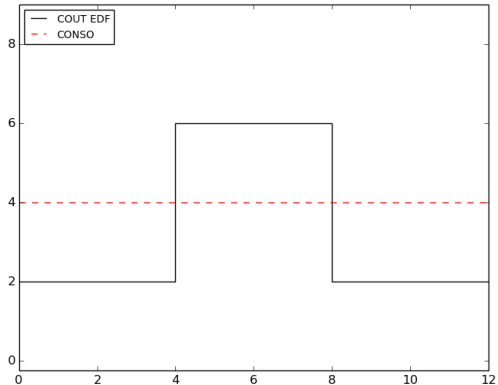
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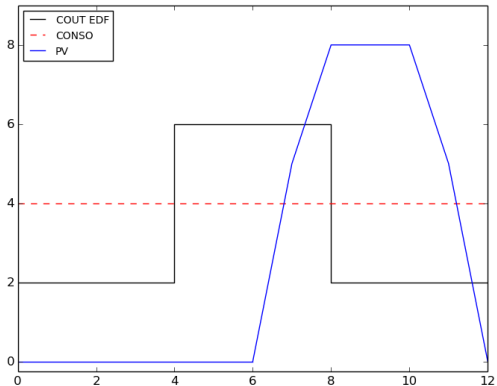
Illustrative example



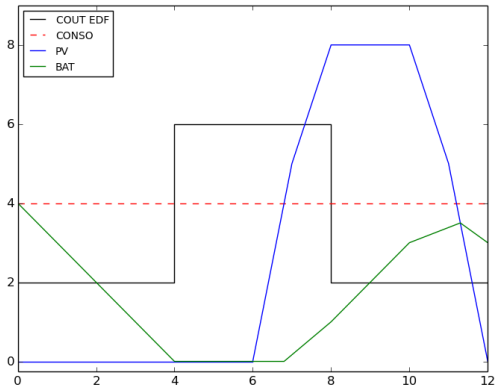
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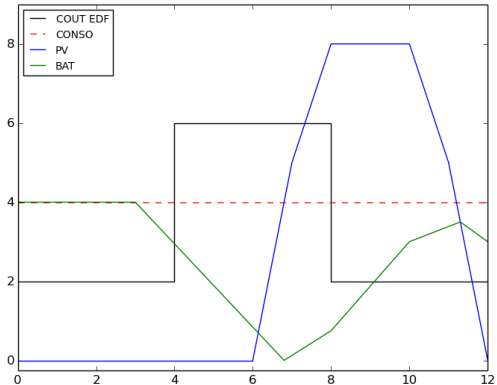
Illustrative example



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Illustrative example



Thank you for your attention.