#### ROADEF 2010 Challenge

A large scale energy optimization problem

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9. July 2010

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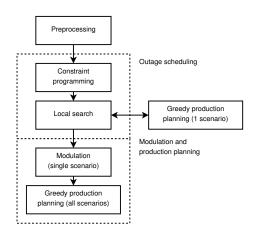
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  - Modulation
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#### Problem introduction

We consider the problem in two parts:

- The outage scheduling part includes:
  - Scheduling constraints (CT13 to CT21) and refueling and fuel level constraints
  - Deciding the number of outages for each type 2 plant
- The production planning part includes:
  - Setting production levels for all scenarios
  - Scenario demand and modulation constraints

#### Solution approach



#### Constraint programming

The CP-solver (made with the Gecode library) is used to find a feasible starting solution for the outage scheduling part.

- Objective: Maximize average online type 2 capacity
- No demand constraints, no scenarios
- Feasible only with respect to scheduling and fuel constraints
- Output: the number of outages, outage start weeks and refueling amounts

Note: in the qualification round we used the ILOG CP-solver, but for the B-instances we had difficulties finding feasible solutions.

#### Constraint programming, cont.

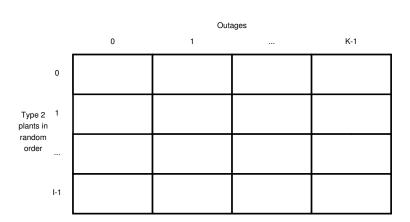
#### Variables:

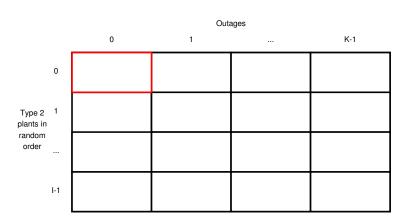
- Outage scheduled, {0,1}
- Outage start week,  $\{0, \ldots, H-1\}$
- Refueling amounts,  $\{R_{min}, \ldots, R_{max}\}$

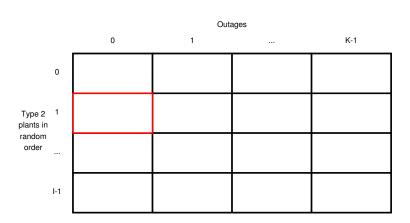
#### Constraints:

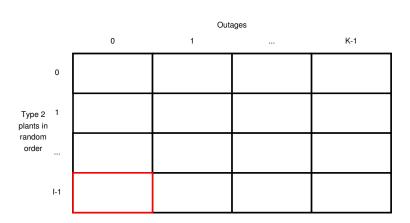
- CT13 to CT21
- Minimum spacing constraints to ensure existence of a feasible refueling schedule

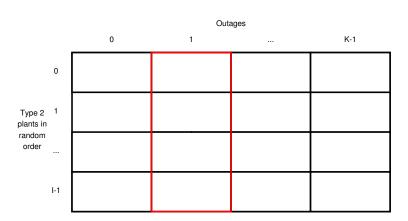
#### Branching strategy













- Outages scheduled: First branch is on scheduled
- 2 Start week: First branch is on earliest week
- 3 Refuel amount: First branch is on maximal refuel amount

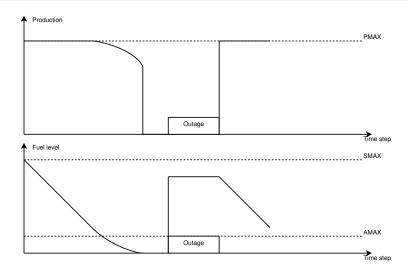
## Greedy production planning

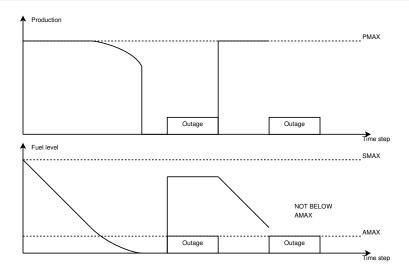
#### For type 2 plants

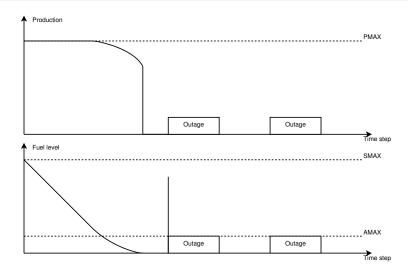
- Find feasible production levels
- Increasing refueling amounts

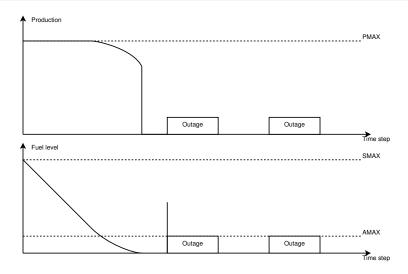
#### For type 1 plants

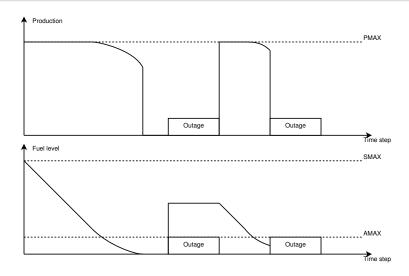
- Cheapest type 1 plant is used first
- Production levels are only set after modulation have been performed and just before writing the final solution









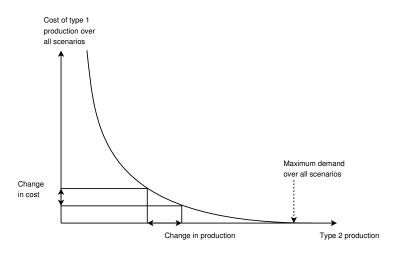


#### Local search

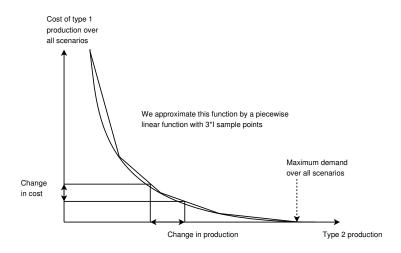
We try to improve on the outage schedule by local search

- Initial solution: We start from the CP-solution
- The neighborhood consist of all possibilities of moving a single outage a few weeks forwards or backwards
  - We only consider moves that are feasible wrt. CT13 to CT21
  - We only consider moving outages less than *m* weeks
- Evaluation: A move is evaluated by the estimated change in production cost. The change consists of changes in
  - Type 2 costs for the affected plant: Re plan production and refuel amounts
  - Type 1 costs: Estimation of the effect over all scenarios.

## Estimating the change in type 1 cost



#### Estimating the change in type 1 cost



#### This approximation to type 1 costs is relatively good:

- In our experience it is correct up to 3 or 4 significant digits
- Evaluating the approximation is a constant time operation, since we maintain the total type 2 production for every time step

#### But somewhat memory expensive:

Need to store an approximation for each time step

#### Meta heuristic, simulated annealing

To guide the local search we use a simple simulated annealing procedure

- Start temperature that gives an acceptance ratio of approximately 0.5
- Exponential cooling with plateaus
- Restart after n idle iterations

#### Modulation strategy

To make the solution feasible we modulate the type 2 power plants according to the minimum demand scenario.

• Idea: Modulate on the type 2 plant which has the shortest time to the next outage.

It is better to modulation per scenario, as we will see now.

#### Results

| Instance    | Results (competition)  | Results (per sc. mod.) |
|-------------|------------------------|------------------------|
| dataB6.txt  | $8.5511 \cdot 10^{10}$ | $8.5544 \cdot 10^{10}$ |
| dataB7.txt  | $8.1900 \cdot 10^{10}$ | $8.1912 \cdot 10^{10}$ |
| dataB8.txt  | $8.3469 \cdot 10^{10}$ | $8.2810 \cdot 10^{10}$ |
| dataB9.txt  | $8.3487 \cdot 10^{10}$ | $8.2851 \cdot 10^{10}$ |
| dataB10.txt | $8.0185 \cdot 10^{10}$ | $7.9150 \cdot 10^{10}$ |

Table: Computational results

The per scenario results are for instances B8, B9 and B10 about 0.8%, 0.8% and 1.2% better respectively.

Problem introduction Solution approach Results Questions

# Thank you for your attention!

Problem introduction Solution approach Results Questions

# Questions?