

Integrated planning of biomass inventory and energy production

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July 5th, 2012

Outline

1. Production planning and investment evaluation
Changing fuel: From coal to wood pellets
2. Mathematical model
Mixed integer linear programming model
3. Benders decomposition
Benders optimality cuts
Handling multiple scenarios
4. Results

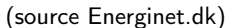
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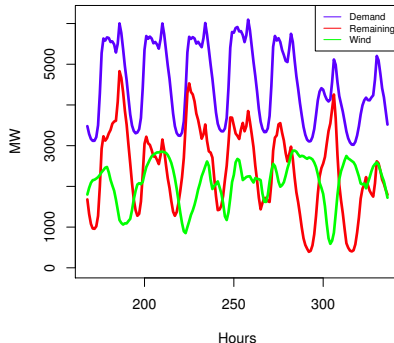


Danish energy system



(source Energinet.dk)

Weekly demand profile



Energy production

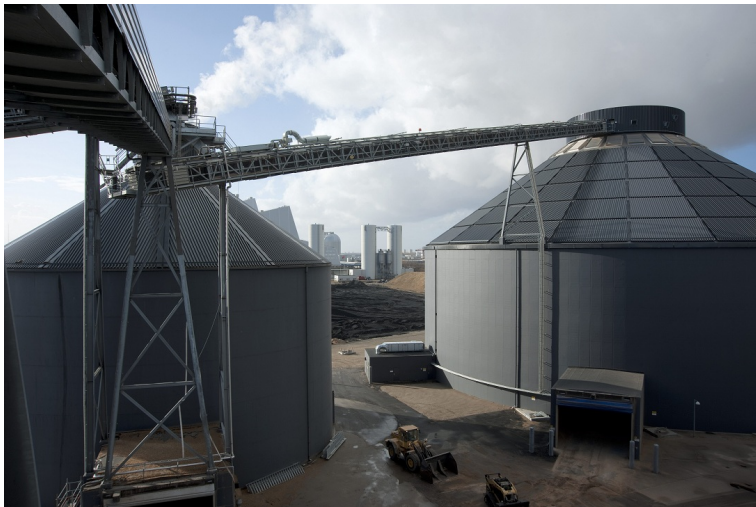
- ▶ Uncontrollable:
 - ▶ Wind power
 - ▶ Solar power
- ▶ Controllable
 - ▶ **Thermal units:**
Providing heat to the local heating area
 - ▶ Connections to neighboring countries
- ▶ Other sources:
 - ▶ SmartGrid
 - ▶ Electric cars



Overview of Avedøre power plant



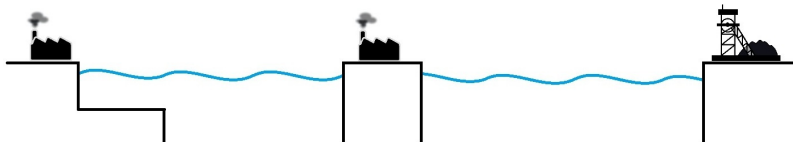
Wood pellet storage at Avedøre



Fuel delivery processes

Logistics differences

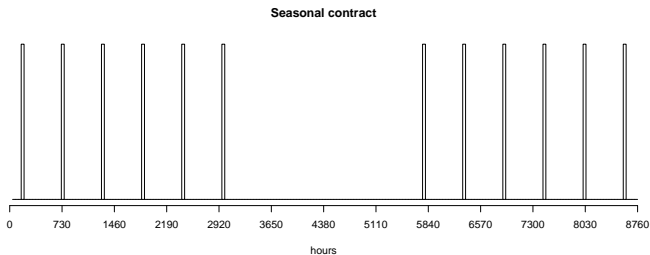
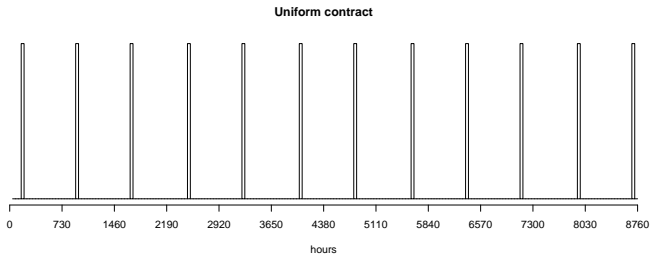
Coal logistics



Wood pellets logistics



Biomass contracts



Two stage stochastic approach

- ▶ Biomass contracts must be decided a year ahead.
- ▶ But future demand, prices and exact delivery times are unknown
 ↪ uncertainty.

Two stage stochastic approach (look-ahead policy):

- ▶ **First stage:** long term decisions on biomass contracts might yield:
 - ▶ Running out of fuel (underflow)
 - ▶ Running out of storage space (overflow)
- ▶ **Second stage:** optimize when uncertainty is revealed
 - ▶ Production of electricity and heat.
 - ▶ Foreign trade (only electricity).
 - ▶ Using an alternative (fossil) fuel
 - ▶ Redirection of deliveries.

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Mixed integer linear programming model

Several scenarios for future uncertainty

Objective function (minimize):

- ▶ Cost of biomass contracts
- ▶ Use of fossil fuel
- ▶ Foreign trade
- ▶ Over/under production (slack/surplus demand)

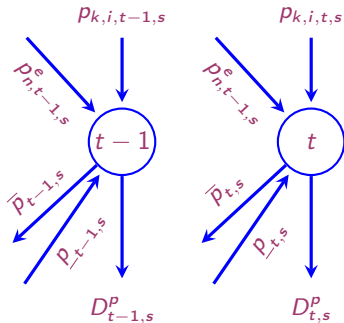
Constraints:

- ▶ Electricity and heat demand
- ▶ Power plant production (including trade with neighboring countries)
- ▶ Biomass fuel levels and redirection of deliveries

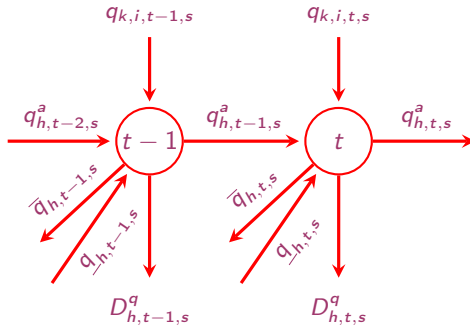
Constraints

Electricity and heat balance

Electricity

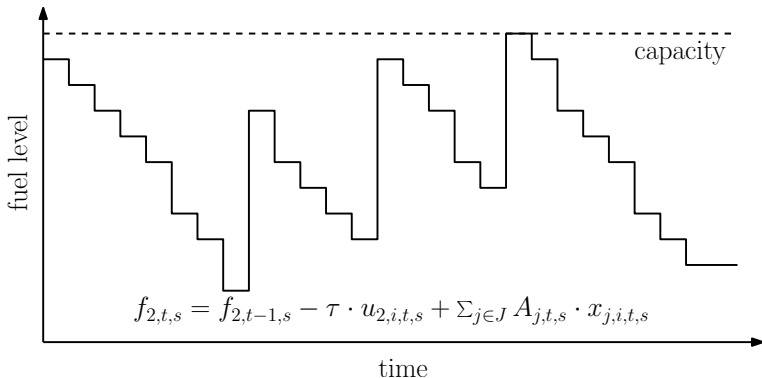


Heat



Constraints

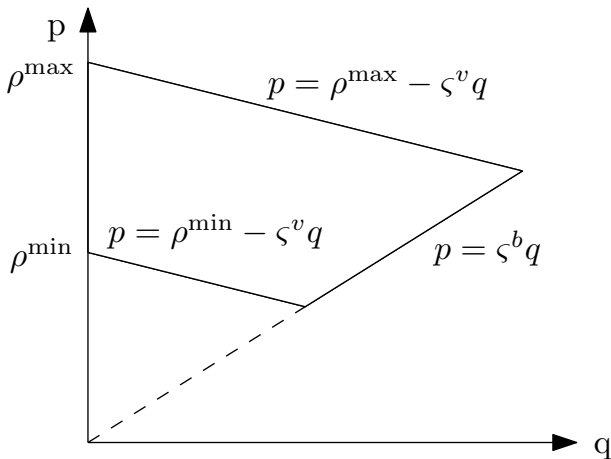
Biomass fuel level constraints



Constraints

Modeling power plant production

Cogeneration power plant



The full model

$$\begin{aligned}
 \min \quad & \sum_{j \in J} C_j^p \cdot w_j + && \text{biomass purchase costs (1a)} \\
 & \frac{1}{|S|} \left(\sum_{i \in I} \sum_{t \in T} \sum_{s \in S} C_i^p \cdot y_{i,t,s} \right) + && \text{unit starting costs (1b)} \\
 & \frac{1}{|S|} \left(\sum_{i \in I} \sum_{t \in T} \sum_{s \in S} \tau \cdot C_{i,s}^u \cdot u_{i,t,s} \right) + && \text{fossil fuel utilization costs (1c)} \\
 & \frac{1}{|S|} \left(\sum_{i \in I} \sum_{t \in T} \sum_{s \in S} \tau \cdot C^q \cdot q_{i,t,s} \right) + && \text{heating production tax costs (1d)} \\
 & \frac{1}{|S|} \left(\sum_{i \in I} \sum_{s \in S} (C^p \cdot \bar{p}_{i,s} + C^E \cdot \underline{p}_{i,s}) \right) + && \text{under/over power costs (1e)} \\
 & \frac{1}{|S|} \left(\sum_{h \in H} \sum_{t \in T} \sum_{s \in S} (C^H \cdot \bar{q}_{h,t,s} + C^S \cdot g_{h,t,s}) \right) + && \text{under/over heat costs (1f)} \\
 & \frac{1}{|S|} \left(\sum_{n \in N} \sum_{t \in T} \sum_{s \in S} \tau \cdot C_{n,t,s}^e \cdot p_{n,t,s}^e \right) && \text{power exchange value (1g)} \\
 \\
 \text{s.t.} \quad & \tau \cdot D_{i,t,s}^p = \sum_{k \in K} \tau \cdot p_{k,i,t,s} - \sum_{n \in N} \tau \cdot p_{n,i,t,s} - \bar{p}_{i,t,s} + \underline{p}_{i,t,s} \quad \forall t \in T, \forall s \in S && \text{power demand (1h)} \\
 & \tau \cdot D_{h,t,s}^h + q_{h,t,s}^a = \sum_{k \in K} \tau \cdot q_{k,h,t,s} - \bar{q}_{h,t,s} + g_{h,t,s} + q_{h,t-1,s}^h \quad \forall h \in H, \forall t \in T, \forall s \in S && \text{heat demand (1i)} \\
 & q_{h,t,s}^a \leq B_{h,t,s}^q \quad \forall h \in H, \forall t \in T, \forall s \in S && \text{heat accumulator (1j)} \\
 & p_{k,i,t,s} + c_i^q \cdot q_{k,i,t,s} \leq p_{i,t,s}^{\max} \cdot v_{k,i,t,s} \quad \forall k \in K, \forall i \in I, \forall t \in T, \forall s \in S && \text{maximum effect (1k)} \\
 & p_{k,i,t,s} + c_i^q \cdot q_{k,i,t,s} \geq p_{i,t,s}^{\min} \cdot v_{k,i,t,s} \quad \forall k \in K, \forall i \in I, \forall t \in T, \forall s \in S && \text{minimum effect (1l)} \\
 & p_{k,i,t,s} \geq c_i^h \cdot q_{k,i,t,s} \quad \forall k \in K, \forall i \in I, \forall t \in T, \forall s \in S && \text{extraction units (1m)} \\
 & p_{k,i,t,s} = c_i^b \cdot q_{k,i,t,s} \quad \forall k \in K, \forall i \in I_b, \forall t \in T, \forall s \in S && \text{back pressure units (1n)} \\
 & q_{k,i,t,s} = 0 \quad \forall k \in K, \forall i \in I_c, \forall t \in T, \forall s \in S && \text{condensing units (1o)} \\
 & e_{i,t,s} \cdot u_{k,i,t,s} = \frac{3.6}{\eta_i} (p_{k,i,t,s} + c_i^q \cdot q_{k,i,t,s}) \quad \forall k \in K, \forall i \in I, \forall t \in T, \forall s \in S && \text{energy production (1p)} \\
 & f_{i,t,s} = f_{i,t-1,s} - \tau \cdot u_{2,i,t,s} + \sum_{j \in J} A_{j,i,t,s} \cdot x_{j,i,t,s} \quad \forall i \in I, \forall t \in T, \forall s \in S && \text{biomass level (1q)} \\
 & f_{i,t,s} \leq B_i^f \quad \forall i \in I, \forall t \in T, \forall s \in S && \text{biomass capacity (1r)} \\
 & p_{n,t,s}^e \leq E_{n,t,s}^{\max} \quad \forall n \in N, \forall t \in T, \forall s \in S && \text{power purchase (1s)} \\
 & p_{n,t,s}^e \geq E_{n,t,s}^{\min} \quad \forall n \in N, \forall t \in T, \forall s \in S && \text{power sale (1t)} \\
 & z_{i,t,s} \geq v_{k,i,t,s} \quad \forall i \in I, \forall t \in T, \forall s \in S && \text{fuel type (1u)} \\
 & y_{i,t,s} \geq z_{i,t,s} - z_{i,t-1,s} \quad \forall i \in I, \forall t \in T, \forall s \in S && \text{plant operation (1v)} \\
 & \sum_{i \in I} x_{j,i,t,s} \leq w_j \quad \forall j \in J, \forall t \in T, \forall s \in S && \text{redirection (1w)} \\
 & p, \bar{p}, \bar{q}, q^h, \bar{q}, \underline{p}, \underline{q}, u, f \in \mathbb{R}^+ && (1x) \\
 & p^e \in \mathbb{R} && (1y) \\
 & w, x, y, z, v \in \mathbb{B} && (1z)
 \end{aligned}$$

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Benders Decomposition

We consider the MILP with *complicating* y -variables, which are the biomass contracts:

$$\min_{x,y} \quad z = c^T x + f^T y$$

$$Ax + By \geq b$$

$$y \in Y$$

$$x \geq 0$$

Benders Decomposition

We consider the MILP with *complicating* y -variables, which are the biomass contracts:

$$\begin{aligned} \min_{x,y} \quad & z = c^T x + f^T y \\ & Ax + By \geq b \\ & y \in Y \\ & x \geq 0 \end{aligned}$$

or emphasizing the two stage approach:

$$\min_{y \in Y} \left[f^T y + \min_{x \geq 0} (c^T x | Ax \geq b - By) \right]$$

Benders optimality cuts

Given a specific set of biomass contracts \bar{y} the dual of the inner problem is:

$$\begin{aligned} \max_u \quad & f^T \bar{y} + (b - B\bar{y})^T u \\ & A^T u \leq c \\ & u \geq 0 \end{aligned}$$

The solution \bar{u} to the dual problem gives a lower bound to the original problem.

The lower bound is valid for all biomass contracts y and the generalization gives:

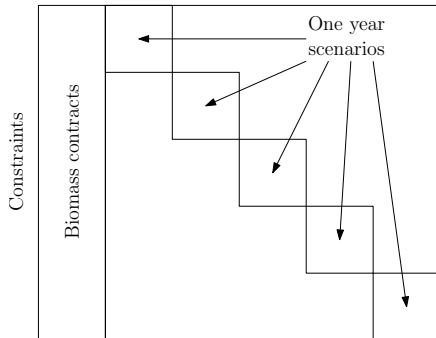
Benders optimality cut

$$z \geq f^T y + (b - By)^T \bar{u}.$$

Handling multiple scenarios

Block angular structure

Variables



Benders optimality cuts for multiple scenarios

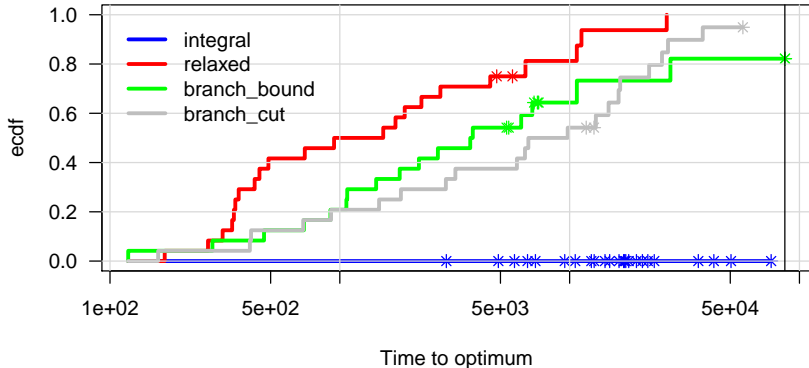
$$\begin{aligned} \min_{y \in Y} \quad & f^T y + \frac{1}{|S|} \sum_{s \in S} z_s \\ \text{s.t.} \quad & z_s \geq (b^s - B^s y)^T u_k^s \\ & \forall s \in S, k = 1 \dots K \end{aligned}$$

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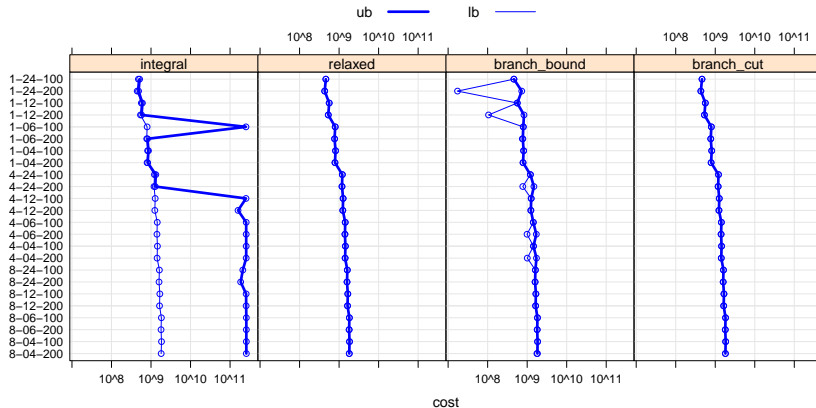
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Time to optimality

Empirical cumulative distribution function of the time to completion of the run for the four models

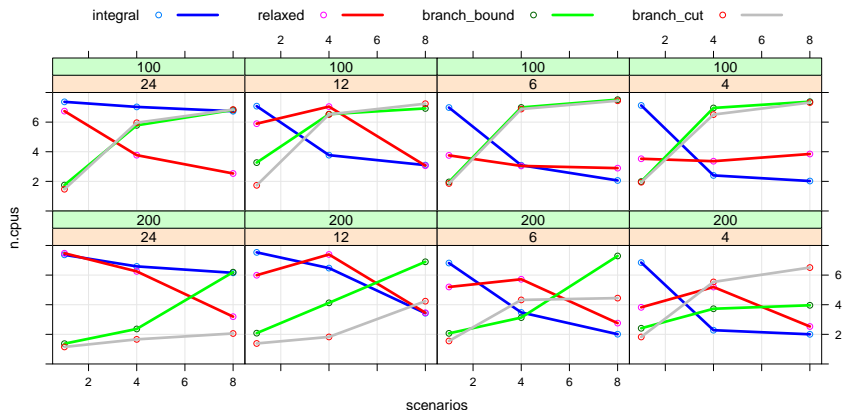


Optimality gap



Exploitation of computational resources

Average number of CPUs used during a run for varying number of scenarios (x-axis) and number of contracts and step size (strip text in the panels).



Conclusions

- ▶ Biomass logistics complicates long term planning
- ▶ Relaxing some of the binary variables does not impact significantly the total cost assessment
- ▶ It is important to consider several scenarios and flexible contracts
- ▶ Benders relaxation does not improve solution times but it is able to exploit computational resources
 - ↪ potential improvement by primal heuristics + more aggressive cuts (future work)

Thank you for your attention!