

# Unit Commitment and Mixed Integer Programming

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- The Unit Commitment Problem
- A Basic Problem and Formulation
- Additional Aspects
- Algorithmic Approaches
  - Lagrangian Relaxation
  - Dynamic Programming
  - MIP
  - Heuristics
- UC and Uncertainty

# The Unit Commitment Problem

UC is the problem of scheduling a park of generators to satisfy the demand for electricity over a certain number of periods at minimum cost/maximum profit.

Classically there is a monopolistic producer.

Recently there are markets for electricity, and a Company owning Generators has to sell their generator capacity on one or more Power Exchange Markets. This leads to both the standard UC problem, as well as other variants.

Generators can be Thermal, Hydro or Renewable (Wind/Solar energy).

Markets lead to price uncertainty, and Renewable leads to production uncertainty.

Time Periods: 5-15-60 minutes, Time Horizon 1 day - 1 week.  
Restrict attention to Thermal: Gas, Coal and Nuclear.

- commodity is nonstorable and instantly perishable
- demand is uncertain (highly correlated with weather)
- demand must be satisfied instantaneously
- very low price elasticity
- transmission (transportation) system is congestion prone
- flows in transmission lines cannot be directly controlled (Kirchoff's law)

Generators  $i = 1, \dots, I$  with lower and upper bounds  $L^i, C^i$  on production per period

Aggregated demand  $d_t$  in period  $t = 1, \dots, T$ .

$f_t^i$  fixed cost if generator  $i$  is active in period  $t$

$q_t^i$  start-up cost if generator  $i$  is turned on in period  $t$

$c_t^i(x)$  cost of producing  $x$  with generator  $i$  in  $t$

## Constraints

Demand Satisfaction

Minimum Up/Down times:  $\alpha^i, \beta^i$

Reserve Constraints to handle Uncertainty in demand - parameter  $\rho$

# Basic Formulation

$x_t^i$  is production of  $i$  in  $t$ ,

$y_t^i = 1$  if generator  $i$  is producing in  $t$ ,  $y_t^i = 0$  otherwise

$$\begin{aligned} \min \quad & \sum_{i,t} (c_t^i(x_t^i) + f_t^i y_t^i + q_t^i (y_t^i - y_{t-1}^i)^+) \\ & L_t^i y_t^i \leq x_t^i \leq C_t^i y_t^i \quad \forall i, t \\ & \sum_i x_t^i = d_t \quad \forall t \\ & \sum_i C^i y_t^i \geq \rho d^t \quad \forall t \\ & y_{t+j}^i \geq y_{t+1}^i - y_t^i \quad j = 1, \dots, \alpha^i, \quad \forall i, t \\ & y_{t+j}^i - y_{t+j-1}^i \leq 1 - y_t^i \quad j = 2, \dots, \beta^i, \quad \forall i, t \\ & x \geq 0, y \in \{0, 1\} \end{aligned}$$

The production cost function  $c_t^i(x)$  is assumed to be

- i) linear
- ii) piecewise linear convex and non-decreasing
- iii) convex quadratic and non-decreasing

## Ramping Constraints

The changes in production level from one period to the next are limited

When on, one cannot increase by more than  $\Delta^i$  and decrease by more than  $\delta^i$ .

When switching on, the production cannot exceed  $\bar{u}^i$ .

When switching off, the production cannot exceed  $\bar{l}^i$ .

## Hot and Cold Starts

# Algorithms? Decomposition Approach

$X$  can be written as

$$X = \left( \prod_{i=1}^I Y^i \right) \cap \left( \prod_{t=1}^n Z^t \right)$$

where  $Y^i =$

$$\begin{aligned} L_t^i y_t^i &\leq x_t^i \leq C_t^i y_t^i \quad \forall i, t \\ y_{t+j}^i &\geq y_{t+1}^i - y_t^i \quad j = 1, \dots, \alpha^i, \forall i, t \\ y_{t+j}^i - y_{t+j-1}^i &\leq 1 - y_{t+1}^i \quad j = 1, \dots, \alpha^i, \forall i, t \\ x^i &\geq 0, y^i \in \{0, 1\} \end{aligned}$$

Is it easy to optimize over  $Y^i$ , or to solve the separation problem for  $Y^i$ /describe  $\text{conv}(Y^i)$ ?



If  $\text{OPT}(Y^i)$  is easy,

Lagrangian Relaxation or Column Generation (Dantzig-Wolfe Decomposition)

If  $\text{SEP}(Y^i)$  is easy (or good approximation of convex hull),

MIP with tightened a priori formulation and/or cutting planes

Single Generator  $i$ . No ramping. Drop index  $i$ .

$$\min \sum_t (F_t(x_t) + f_t y_t + q_t (y_t - y_{t-1})^+)$$

$$L y_t \leq x \leq C y_t \quad \forall t$$

*Min Up/Down Times*

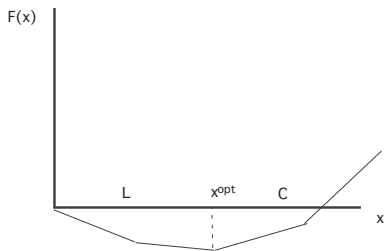
$$x \geq 0, y \in \{0, 1\}$$

# Two Step Optimization: Step 1

Step 1: If  $y_t = 1$ , then

$$x_t^{opt} = \arg \min \{ F(x) : L \leq x \leq C \}$$

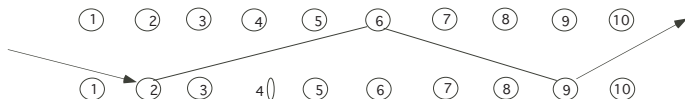
where  $F(x) = c(x) - \pi x$  is of the form



# Two Step Optimization: Step 2

If the generator is on from periods  $k$  to  $t - 1$  and off in  $k - 1$  and  $t$  with  $t - k \geq \alpha$ , then upward arc  $(k, t)$  and the associated cost is

$$q_k + \sum_{u=k}^{t-1} (F_u(x_u^{opt}) + f_u).$$



On from periods 2-5. Off in period 1 and periods 6-8.

Cost of upper arc 2-5:  $F^2(x_2) + f_2 + \dots + F_5(x_5) + f_5 + g_2$

Suppose that  $\alpha = 2$  and  $\beta = 3$ .

# Lagrangian relaxation

$$\begin{aligned} \min \sum_{i,t} (c_t^i(x_t^i) + f_t^i y_t^i + q_t^i (y_t^i - y_{t-1}^i)^+) \\ \sum_i x_t^i = d_t \quad \forall t \\ \sum_i C^i y_t^i \geq \rho d_t \quad \forall t \\ (x^i, y^i) \in Y^i \quad \forall i \end{aligned}$$

Dualizing the demand and reserve constraints gives the Lagrangian dual:

$$L(\pi, \mu) =$$

$$\begin{aligned} \min \sum_{i,t} (c_t^i(x_t^i) + f_t^i y_t^i + q_t^i (y_t^i - y_{t-1}^i)^+) + \sum_t \pi_t (d_t - \sum_i x_t^i) + \sum_t \mu_t (\rho d_t - \sum_i C^i y_t^i) \\ (x^i, y^i) \in Y^i \quad \forall i \end{aligned}$$

$$L(\pi, \mu) = \min \sum_{i,t} ((f_t^i - \mu_t C^i) y_t^i + q_t^i (y_t^i - y_{t-1}^i)^+ + c_t^i (x_t^i) - \pi^i x_t^i) \\ + \sum_t (\pi_t d_t + \mu_t \rho d_t) : \quad (x^i, y^i) \in Y^i \forall i.$$

$$L(\pi, \mu) = \sum_i L^i(\pi, \mu) + \sum_t (\pi_t d_t + \mu_t \rho d_t)$$

$$L^i(\pi, \mu) = \min \sum_t ((f_t^i - \mu_t C^i) y_t^i + q_t^i (y_t^i - y_{t-1}^i)^+ + c_t^i (x_t^i) - \pi^i x_t^i) \\ (x^i, y^i) \in Y^i.$$

# Lower Bounds from Lagrangean Dual

$$Z \geq L(\pi, \mu) \quad \forall \pi \in \mathbb{R}_+^T, \mu \in \mathbb{R}^t.$$

Lagrangean Dual

$$W = \max\{L(\pi, \mu) : \pi \in \mathbb{R}_+^T, \mu \in \mathbb{R}^t\}$$

Algorithms to solve Lagrangean Dual i.e. find optimal multipliers  $(\pi, \mu)$

Subgradient Algorithms

Bundle Algorithms

Dantzig-Wolfe / Column Generation

# Feasible Solutions and Upper Bounds

Lagrangian relaxation provides solutions  $(x, y)$  at each iteration that satisfy the generator constraints, but can violate the demand constraints.

The bundle or column generation algorithms provide convexified solution at each iteration that satisfies all the constraints

Combinatorial Heuristic:

Run at every step of the iterative algorithm

- Fix hydro power (if any) as in convexified solution
- Greedy-type heuristic selecting generators so that sufficient capacity to satisfy remaining demand. Interpret fractional  $y$  as probabilities.
- Given  $y$  fixed, solve the economic dispatch problem. Large-scale convex problem. Not guaranteed to be feasible.



# Extensions: Ramping and Convex Production Costs

Construct a digraph with source  $s$ , sink  $d$  and nodes  $(h, k)$  representing switch-on in  $h$  and generator off on  $k$   
Arc  $(h, k)$  to  $(r, q)$  if it is feasible to turn on in  $r$  after turning off in  $k$ . Cost that of a start-up in  $r$ .  
Arc  $s$  to  $(h, k)$  if possible initial state.  
Node costs: Solve Economic Dispatch for  $(h, k)$

$$z_{hk} = \min \sum_i c^i(x_t^i) : (x^i, y^i, z^i) \in Y^i, y_t^i = 1 \quad t = h, \dots, k-1, \text{ else } y_t^i = 0$$

Then solve min cost path problem.  $O(n^2)$  nodes and  $O(n^4)$  arcs.

With care, can calculate all  $z_{hk}$  and solve shortest path problem in  $O(T^3)$ .

Table: Solving the Lagrangean Dual: no hydro units

I	secs	iter	sol	% gap
20	8	189	34	0.44
50	17	195	33	0.26
100	46	213	21	0.48
200	134	317	67	0.06

# MIP: Tightening Formulations

## Min Up-Down Times

$$\begin{aligned}y_{t+j}^i &\geq y_{t+1}^i - y_t^i \quad j = 1, \dots, \alpha^i, \forall i, t \\y_{t+j}^i - y_{t+j-1}^i &\leq 1 - y_{t+1}^i \quad j = 1, \dots, \alpha^i, \forall i, t \\y^i &\in \{0, 1\}\end{aligned}$$

With start-up variables:

$$\begin{aligned}z_t &\geq y_t - y_{t-1}, z_t \leq y_t, z_t \leq 1 - y_{t-1} \quad \forall t \\ \sum_{j=1}^{\alpha} z_{t+j} &\leq y_{t+\alpha} \quad \forall t \\ \sum_{j=1}^{\beta} z_{t+j} &\leq 1 - y_t \quad \forall t\end{aligned}$$

Without start-up variables: using  $y_{j+1} - y_j \leq z_{j+1}$  and  $0 \leq z_{j+1}$ , we obtain

$$\sum_{j \in S \cap [t+1, t+\alpha]} (y_j - y_{j-1}) \leq y_{t+\alpha}.$$

Up-Ramping.

$$\begin{aligned}Ly_t &\leq x_t \leq Cy_t \quad \forall t \\x_{t+1} - x_t &\leq \Delta_t y_t + u_{t+1}(1 - y_{t+1}) \quad \forall t \\y &\in \{0, 1\}^T\end{aligned}$$

Valid inequalities approximating convex hull.

$$\begin{aligned}x_{t+1} - x_t &\leq (\bar{u} - L - \Delta)z_{t+1} + (L + \Delta)y_{t+1} - Ly_t \\x_t &\leq Cy_t - (C - \bar{u})z_t\end{aligned}$$

In any extreme point, each  $x_t$  takes one of  $K$  values  $p_1, \dots, p_K$  where  $K$  is  $O(T)$ .

OPT as a shortest path problem with nodes  $(t, p_k)$ ,

Arc variable  $\alpha_{j,k,t}$  from  $(j, t-1)$  to  $(k, t)$  with cost  $f_t + c_t(k)$

Arc variable  $\beta_{k,t}$  from  $(0, t-1)$  to  $(k, t)$  with cost  $g_t + f_t + c_t(k)$

Extended formulation:

$$x_t = \sum_j k \alpha_{jkt} + \sum_k k \beta_{kt}$$

$$y_t = \sum_j \sum_{k \neq 0} \alpha_{jkt} + \sum_{k \neq 0} \beta_{kt}$$

$$z_t = \sum_{k \neq 0} \beta_{kt}$$

Relax and Fix  
Local Re-optimization  
LP Fixing

# Uncertain Unit Commitment

- Scenario Tree Approaches  
Select a certain number of demand/price scenarios  $k = 1, \dots, K$ . Decide generator variables  $y, z$ , observe outcomes and then decide production levels  $x^k, k = 1, \dots, K$ .  
Very large MIPs. Possible too use Benders' decomposition
- Robust Optimization  
Demand/price scenarios lie in a "small" polytope. Solve for all possible "worst-case" outcomes in the polytope.
- Chance-Constrained Optimization Satisfy "demand constraints" with 99% probability.



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