

MOPEC: Multiple Optimization Problems with Equilibrium Constraints

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The PIES Model (Hogan)

$$\begin{aligned} \min_x \quad & c^T x && \text{cost} \\ \text{s.t.} \quad & Ax = d(p) && \text{balance} \\ & Bx = b && \text{technical constr} \\ & x \geq 0 && \end{aligned}$$

- Issue is that p is the multiplier on the “balance” constraint of LP
- Extended Mathematical Programming (EMP) facilitates annotations of models to describe additional structure
- Can solve the problem by writing down the KKT conditions of this LP, forming an LCP and exposing p to the model
- EMP does this automatically from the annotations

Reformulation details

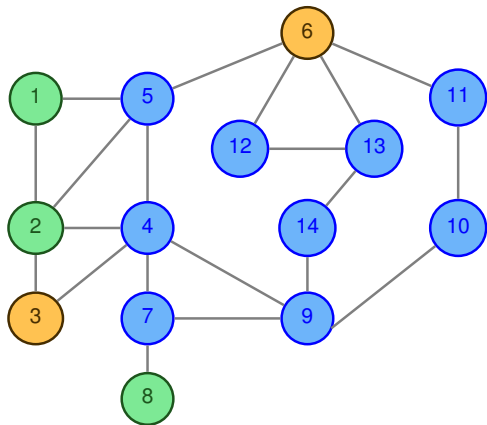
$$\begin{array}{lcl} 0 = Ax - d(p) & \perp & \mu \\ 0 = Bx - b & \perp & \lambda \\ 0 \leq -A^T \mu - B^T \lambda + c & \perp & x \geq 0 \end{array}$$

- **empinfo: dualvar p balance**
- replaces $\mu \equiv p$
- LCP/MCP is then solvable using PATH

$$z = \begin{bmatrix} p \\ \lambda \\ x \end{bmatrix}, \quad F(z) = \begin{bmatrix} A \\ B \\ -A^T & -B^T \end{bmatrix} \begin{bmatrix} p \\ \lambda \\ x \end{bmatrix} + \begin{bmatrix} -d(p) \\ -b \\ c \end{bmatrix}$$

Transmission Line Expansion Model (F./Tang)

$$\min_{x \in X} \sum_{\omega} \pi_{\omega} \sum_{i \in N} d_i^{\omega} p_i^{\omega}(x)$$



- Nonlinear system to describe power flows over (large) network
- Multiple time scales
- Dynamics (bidding, failures, ramping, etc)
- Uncertainty (demand, weather, expansion, etc)
- $p_i^{\omega}(x)$: Price (LMP) at i in scenario ω as a function of x
- Use other models to construct approximation of $p_i^{\omega}(x)$

Solution approach

- Use derivative free method for the upper level problem (1)
- Requires $p_i^\omega(x)$
- Construct these as multipliers on demand equation (per scenario) in an Economic Dispatch (market clearing) model
- But transmission line capacity expansion typically leads to generator expansion, which interacts directly with market clearing
- Interface blue and black models using Nash Equilibria (as EMP):

empinfo: equilibrium

forall f: min expcost(f) y(f) budget(f)

forall ω : min scencost(ω) q(ω) ...

Generator Expansion (2): $\forall f \in F$:

$$\min_{y_f} \sum_{\omega} \pi_{\omega} \sum_{j \in G_f} C_j(y_j, q_j^{\omega}) - r(h_f - \sum_{j \in G_f} y_j)$$

s.t. $\sum_{j \in G_f} y_j \leq h_f, y_f \geq 0$

G_f : Generators of firm $f \in F$
 y_j : Investment in generator j
 q_j^{ω} : Power generated at bus j in scenario ω
 C_j : Cost function for generator j
 r : Interest rate

Market Clearing Model (3): $\forall \omega$:

$$\min_{z, \theta, q^{\omega}} \sum_f \sum_{j \in G_f} C_j(y_j, q_j^{\omega}) \quad \text{s.t.}$$

$$q_j^{\omega} - \sum_{i \in I(j)} z_{ij} = d_j^{\omega} \quad \forall j \in N(\perp p_j^{\omega})$$

$$z_{ij} = \Omega_{ij}(\theta_i - \theta_j) \quad \forall (i, j) \in A$$

$$-b_{ij}(x) \leq z_{ij} \leq b_{ij}(x) \quad \forall (i, j) \in A$$

$$\underline{u}_j(y_j) \leq q_j^{\omega} \leq \bar{u}_j(y_j)$$

z_{ij} : Real power flowing along line ij
 q_j^{ω} : Real power generated at bus j in scenario ω
 θ_i : Voltage phase angle at bus i
 Ω_{ij} : Susceptance of line ij
 $b_{ij}(x)$: Line capacity as a function of x
 $\underline{u}_j(y)$, $\bar{u}_j(y)$: Generator j limits as a function of y

MOPEC

$$\min_{x_i} \theta_i(x_i, x_{-i}, y) \text{ s.t. } g_i(x_i, x_{-i}, y) \leq 0, \forall i$$

equilibrium

min theta(1) x(1) g(1)

...

min theta(m) x(m) g(m)

is solved in a Nash manner

- Allows multipliers from one problem to be used in another problems
dualvar p g(1)

Feasibility

$$\text{KKT of } \min_{y_f \in Y} \sum_{\omega} \pi_{\omega} \sum_{j \in G_f} C_j(y_j, q_j^{\omega}) - r(h_f - \sum_{j \in G_f} y_j) \quad \forall f \in F \quad (2)$$

$$\text{KKT of } \min_{(z, \theta, q^{\omega}) \in Z(x, y)} \sum_f \sum_{j \in G_f} C_j(y_j, q_j^{\omega}) \quad \forall \omega \quad (3)$$

- Models (2) and (3) form a complementarity problem (CP via EMP)
- Solve (3) as NLP using global solver (actual $C_j(y_j, q_j^{\omega})$ are not convex), per scenario (SNLP) this provides starting point for CP
- Solve (KKT(2) + KKT(3)) using EMP and PATH, then repeat
- Identifies CP solution whose components solve the scenario NLP's (3) to global optimality

Scenario	ω_1	ω_2
Probability	0.5	0.5
Demand Multiplier	8	5.5

SNLP (1):

Scenario	q_1	q_2	q_3	q_6	q_8
ω_1	3.05	4.25	3.93	4.34	3.39
ω_2		4.41	4.07	4.55	

Scenario	ω_1	ω_2
Probability	0.5	0.5
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SNLP (1):

Scenario	q_1	q_2	q_3	q_6	q_8
ω_1	3.05	4.25	3.93	4.34	3.39
ω_2		4.41	4.07	4.55	

EMP (1):

Scenario	q_1	q_2	q_3	q_6	q_8
ω_1	2.86	4.60	4.00	4.12	3.38
ω_2		4.70	4.09	4.24	

Firm	y_1	y_2	y_3	y_6	y_8
f_1	167.83	565.31			266.86
f_2			292.11	207.89	

Scenario	ω_1	ω_2
Probability	0.5	0.5
Demand Multiplier	8	5.5

SNLP (2):

Scenario	q_1	q_2	q_3	q_6	q_8
ω_1	0.00	5.35	4.66	5.04	3.91
ω_2		4.70	4.09	4.24	

Scenario	ω_1	ω_2
Probability	0.5	0.5
Demand Multiplier	8	5.5

SNLP (2):

Scenario	q_1	q_2	q_3	q_6	q_8
ω_1	0.00	5.35	4.66	5.04	3.91
ω_2		4.70	4.09	4.24	

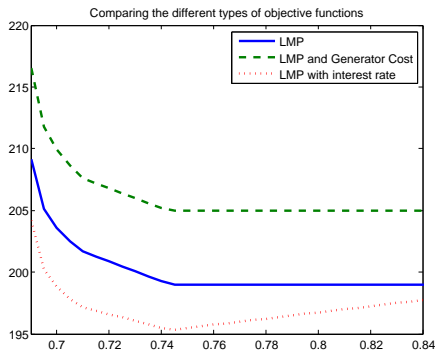
EMP (2):

Scenario	q_1	q_2	q_3	q_6	q_8
ω_1	0.00	5.34	4.62	5.01	3.99
ω_2		4.71	4.07	4.25	

Firm	y_1	y_2	y_3	y_6	y_8
f_1	0.00	622.02			377.98
f_2			283.22	216.79	

Observations

- But this is simply one function evaluation for the outer “transmission capacity expansion” problem
- Number of critical arcs typically very small
- But in this case, p_j^ω are very volatile
- Outer problem is small scale, objectives are open to debate, possibly ill conditioned
- Economic dispatch should use AC power flow model
- Structure of market open to debate
- Types of “generator expansion” also subject to debate
- Suite of tools is very effective in such situations



EMP: variational inequalities

Allows (GAMS) models to be manipulated to form other problems of interest via a simple EMP info file:

- $VI(f, C)$:

$$0 \in f(x) + N_C(x)$$

vi f x cons

generates a variational inequality where C defined by 'cons'

- Either generates the equivalent complementarity (KKT) problem, or provides problem structure for algorithmic exploitation
- Extension of (square) nonlinear systems and mixed complementarity problems
- QVI can be specified in the same manner

MOPEC

$$\min_{x_i} \theta_i(x_i, x_{-i}, p) \text{ s.t. } g_i(x_i, x_{-i}, p) \leq 0, \forall i$$

and

$$p \text{ solves VI}(h(x, \cdot), C)$$

equilibrium

min theta(1) x(1) g(1)

...

min theta(m) x(m) g(m)

vi h p cons

is solved in a Nash manner

MOPEC

$$\min_{x_i} \theta_i(x_i, x_{-i}, p) \text{ s.t. } g_i(x_i, x_{-i}, p) \leq 0, \forall i$$

and

$$h(x, p) = 0$$

equilibrium

min theta(1) x(1) g(1)

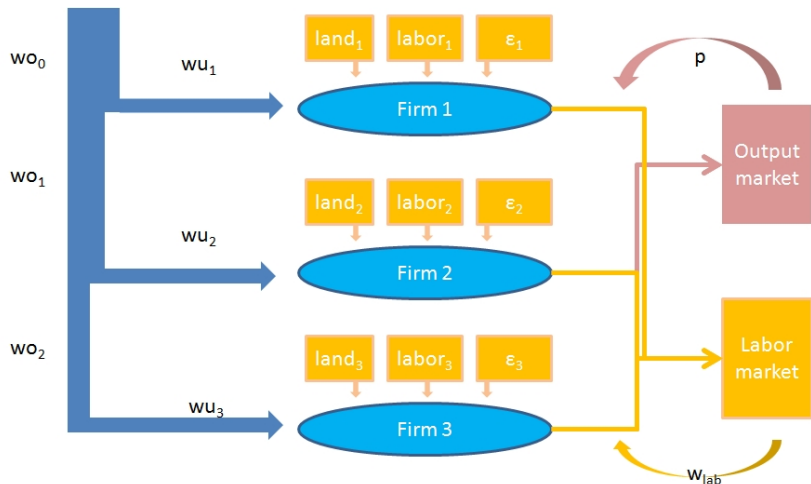
...

min theta(m) x(m) g(m)

vi h p cons

is solved in a Nash manner

Water rights pricing (Britz/F./Kuhn)



The model AO

$$\begin{aligned} \max_{q_i, x_i, w_o_i \geq 0} & \sum_i \left(q_i \cdot p - \sum_{f \in \{int, lab\}} x_{i,f} \cdot w_f \right) \\ \text{s.t.} & q_i \leq \prod_f (x_{i,f} + e_{i,f})^{\epsilon_{i,f}} \\ & x_{i,land} \leq e_{i,land} \\ & w_{o_{i-1}} = x_{i,wat} + w_{o_i} \end{aligned}$$

$$\begin{aligned} 0 \leq \sum_i q_i - d(p) & \perp p \geq 0 \\ 0 \leq \sum_i e_{i,lab} - \sum_i x_{i,lab} & \perp w_{lab} \geq 0 \end{aligned}$$

The model IO

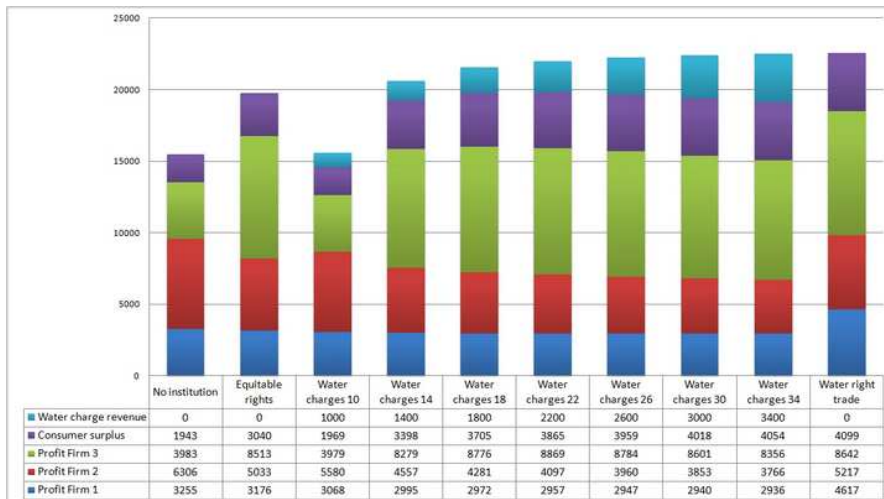
$$\begin{aligned} \max_{q_i, x_{i,f}, w_{oj} \geq 0} & \quad \left(q_i \cdot p - \sum_f x_{i,f} \cdot w_f \right) \\ \text{s.t.} & \quad q_i \leq \prod_f (x_{i,f} + e_{i,f})^{e_{i,f}} \\ & \quad x_{i,land} \leq e_{i,land} \\ & \quad w_{oj-1} = x_{i,wat} + w_{oj} \\ \\ 0 \leq & \quad \sum_i q_i - d(p) \quad \perp \quad p \geq 0 \\ 0 \leq & \quad \sum_i e_{i,lab} - \sum_i x_{i,lab} \quad \perp \quad w_{lab} \geq 0 \end{aligned}$$

The model IO

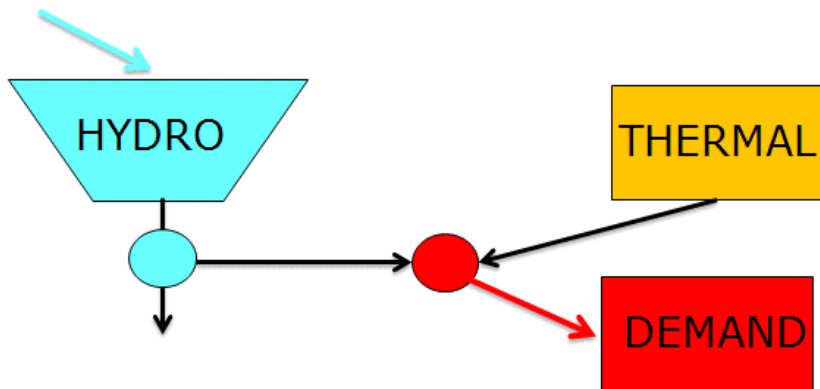
$$\begin{aligned}
 & \max_{q_i, x_i, w_{o_i}, w_{r_i}^b, w_{r_i}^s \geq 0} \left(q_i \cdot p - \sum_f x_{i,f} \cdot w_f - w_{r_i}^b \cdot (w_{wr} + \tau) + w_{r_i}^s \cdot w_{wr} \right) \\
 \text{s.t.} \quad & q_i \leq \prod_f (x_{i,f} + e_{i,f})^{\epsilon_{i,f}} \\
 & x_{i,land} \leq e_{i,land} \\
 & w_{o_{i-1}} = x_{i,wat} + w_{o_i} \\
 & w_{r_i} + w_{r_i}^b \geq x_{i,wat} + w_{r_i}^s
 \end{aligned}$$

$$\begin{aligned}
 0 &\leq \sum_i q_i - d(p) \quad \perp \quad p \geq 0 \\
 0 &\leq \sum_i e_{i,lab} - \sum_i x_{i,lab} \quad \perp \quad w_{lab} \geq 0 \\
 0 &\leq \sum_i w_{r_i}^s - \sum_i w_{r_i}^b \quad \perp \quad w_{wr} \geq 0
 \end{aligned}$$

Different Management Strategies



Hydro-Thermal System (Philpott/F./Wets)



Simple electricity system optimization problem

$$\text{SSP: } \min \sum_{j \in \mathcal{T}} C_j(v(j)) - \sum_{i \in \mathcal{H}} V_i(x(i))$$

$$\begin{aligned} \text{s.t. } & \sum_{i \in \mathcal{H}} U_i(u(i)) + \sum_{j \in \mathcal{T}} v(j) \geq d, \\ & x(i) = x_0(i) - u(i), \quad i \in \mathcal{H} \\ & u(i), v(j), x(i) \geq 0. \end{aligned}$$

- $u(i)$ water release of hydro reservoir $i \in \mathcal{H}$
- $v(j)$ thermal generation of plant $j \in \mathcal{T}$
- production function U_i (strictly concave) converts water release to energy
- water level reservoir $i \in \mathcal{H}$ is denoted $x(i)$
- $C_j(v(j))$ denote the cost of generation by thermal plant
- $V_i(x(i))$ to be the future value of terminating the period with storage x (assumed separable)

SSP equivalent to CE

Thermal plants solve

$$\text{TP}(j): \quad \max \quad \pi v(j) - C_j(v(j))$$

$$\text{s.t.} \quad v_1(j) \geq 0.$$

The hydro plants $i \in \mathcal{H}$ solve

$$\text{HP}(i): \quad \max \quad \pi U_i(u(i)) + V_i(x(i))$$

$$\text{s.t.} \quad x(i) = x_0(i) - u(i) \\ u(i), x(i) \geq 0.$$

Perfectly competitive (Walrasian) equilibrium is a MOPEC

$$\begin{aligned} \text{CE:} \quad & u(i), x(i) \in \arg \max \text{HP}(i), & i \in \mathcal{H}, \\ & v(j) \in \arg \max \text{TP}(j), & j \in \mathcal{T}, \\ & 0 \leq (\sum_{i \in \mathcal{H}} U_i(u(i)) + \sum_{j \in \mathcal{T}} v(j)) - d \perp \pi \geq 0, \end{aligned}$$

GAMS/EMP: Stochastic programming tools

- GAMS has extended mathematical programming tools to build “models of models”
- Given the core model, can annotate parameters as “random variables”
- Automatically solves expected value problem
- Can solve using deterministic equivalent or specialized solvers (including Bender’s decomposition, importance sampling (DECIS), etc)
- Also allows for a variety of new constructs (such as risk measures and chance constraints)

$$\mathbb{R}_\omega \left[c^0(x) + \sum_{t=0}^T p_{\omega t} (q_{\omega t}^+ - q_{\omega t}^-) + c^1(q_{\omega t}^+ + q_{\omega t}^-) \right]$$

Two stage problems

$$\text{TP}(j): \quad \max \quad \pi_1 v_1(j) - C_j(v_1(j)) + \\ R_\omega[\pi_2(\omega)v_2(j, \omega) - C_j(v_2(j, \omega))]$$

$$\text{s.t.} \quad v_1(j) \geq 0, \quad v_2(j, \omega) \geq 0, \quad \text{for all } \omega \in \Omega.$$

$$\text{HP}(i): \quad \max \quad \pi_1 U_i(u_1(i)) + \\ R_\omega[\pi_2(\omega)U_i(u_2(i, \omega)) + V_i(x_2(i, \omega))]$$

$$\text{s.t.} \quad x_1(i) = x_0(i) - u_1(i) + h_1(i), \\ x_2(i, \omega) = x_1(i) - u_2(i, \omega) + h_2(i, \omega), \quad \text{for all } \omega \in \Omega, \\ u_1(i), x_1(i) \geq 0, \quad u_2(i, \omega), x_2(i, \omega) \geq 0, \quad \text{for all } \omega \in \Omega.$$

Results

- Suppose every agent is risk neutral and has knowledge of all deterministic data, as well as sharing the same probability distribution for inflows. **SP solution is same as CE solution**
- Using coherent risk measure (weighted sum of expected value and conditional variance at risk), 10 scenarios for rain
 - 1 High initial storage: risk-averse central plan (**RSP**) and the risk-averse competitive equilibrium (**RCE**) **have same solution** (but different to risk neutral case)
 - 2 Low initial storage: **RSP and RCE are very different**. Since the hydro generator and the system do not agree on a worst-case outcome, the probability distributions that correspond to an equivalent risk neutral decision will not be common.
 - 3 **Extension: Construct MOPEC models for trading risk**

What is EMP?

Annotates existing equations/variables/models for modeler to provide/define additional structure

- equilibrium
- vi (agents can solve min/max/vi)
- bilevel (reformulate as MPEC, or as SOCP)
- disjunction (or other constraint logic primitives)
- randvar
- dualvar (use multipliers from one agent as variables for another)
- extended nonlinear programs (library of plq functions)

Currently available within GAMS

Conclusions

- Optimization helps understand what drives a system
- Collections of models needed for specific decisions
- Modern optimization within applications requires multiple model formats, computational tools and sophisticated solvers
- EMP model type is clear and extensible, additional structure available to solver
- Uncertainty is present everywhere
- We need not only to quantify it, but we need to hedge/control/ameliorate it
- Stochastic MOPEC models capture behavioral effects (as an EMP)
- Policy implications addressable using Stochastic MOPEC
- Extended Mathematical Programming available within the GAMS modeling system