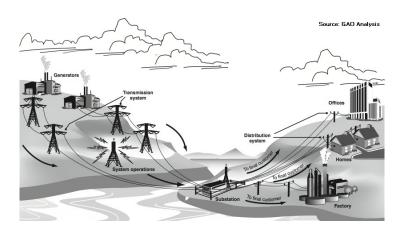
Optimization and Equilibrium in Energy Economics

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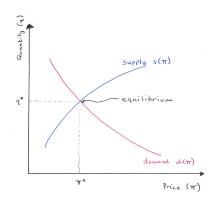
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Power generation, transmission and distribution



- Determine generators' output to reliably meet the load
 - ▶ \sum Gen MW $\geq \sum$ Load MW, at all times.
 - ▶ Power flows cannot exceed lines' transfer capacity.

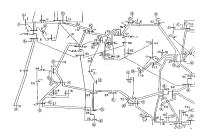
Single market, single good: equilibrium



Walras:
$$0 \le s(\pi) - d(\pi) \perp \pi \ge 0$$

Market design and rules to foster competitive behavior/efficiency

 Spatial extension: Locational Marginal Prices (LMP) at nodes (buses) in the network



- Supply arises often from a generator offer curve (lumpy)
- Technologies and physics affect production and distribution

The PIES Model (Hogan) - Optimal Power Flow (OPF)

$$\min_{x} c(x)$$
 cost
s.t. $Ax \ge q$ balance
 $Bx = b, x \ge 0$ technical constr

The PIES Model (Hogan) - Optimal Power Flow (OPF)

$$\min_{x} c(x)$$
 cost s.t. $Ax \ge d(\pi)$ balance $Bx = b, x \ge 0$ technical constr

- $ullet q=d(\pi)$: issue is that π is the multiplier on the "balance" constraint
- Such multipliers (LMP's) are critical to operation of market
- Can try to solve the problem iteratively (shooting method):

$$\pi^{new} \in \mathsf{multiplier}(\mathit{OPF}(d(\pi)))$$

Alternative: Form KKT of QP, exposing π to modeler

$$0 \le Ax - d(\pi) \qquad \qquad \bot \quad \mu \ge 0$$

$$0 = Bx - b \qquad \qquad \bot \quad \lambda$$

$$0 \le \nabla c(x) - A^{T} \mu - B^{T} \lambda \quad \bot \quad x \ge 0$$

- ullet empinfo: dualvar π balance
- Fixed point: replaces $\mu \equiv \pi$

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- ullet empinfo: dualvar π balance
- Fixed point: replaces $\mu \equiv \pi$
- LCP/MCP is then solvable using PATH

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

- Existence, uniqueness, stability from variational analysis
- EMP does this automatically from the annotations



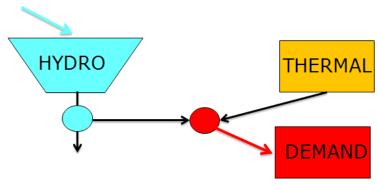
Other applications of complementarity

Complementarity can model fixed points and disjunctions

- Economics: Walrasian equilibrium (supply equals demand), taxes and tariffs, computable general equilibria, option pricing (electricity market), airline overbooking
- Transportation: Wardropian equilibrium (shortest paths), selfish routing, dynamic traffic assignment
- Applied mathematics: Free boundary problems
- Engineering: Optimal control (ELQP)
- Mechanics: Structure design, contact problems (with friction)
- Geology: Earthquake propogation

Good solvers exist for large-scale instances of Complementarity Problems

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



- Competing agents (consumers, or generators in energy market)
- Each agent minimizes objective independently (cost)
- Market prices are function of all agents activities

Simple electricity "system optimization" problem

SO:
$$\max_{d_k, u_i, v_j, x_i \ge 0} \sum_{k \in \mathcal{K}} W_k(d_k) - \sum_{j \in \mathcal{T}} C_j(v_j) + \sum_{i \in \mathcal{H}} V_i(x_i)$$
s.t.
$$\sum_{i \in \mathcal{H}} U_i(u_i) + \sum_{j \in \mathcal{T}} v_j \ge \sum_{k \in \mathcal{K}} d_k,$$

$$x_i = x_i^0 - u_i + h_i^1, \quad i \in \mathcal{H}$$

- u_i water release of hydro reservoir $i \in \mathcal{H}$
- ullet v $_j$ thermal generation of plant $j\in\mathcal{T}$
- x_i water level in reservoir $i \in \mathcal{H}$
- ullet prod fn U_i (strictly concave) converts water release to energy
- \bullet $C_j(v_j)$ denote the cost of generation by thermal plant
- $V_i(x_i)$ future value of terminating with storage x (assumed separable)
- $W_k(d_k)$ utility of consumption d_k



Decomposition by prices π

$$\begin{aligned} \max_{\boldsymbol{d_k},\boldsymbol{u_i},\boldsymbol{v_j},\boldsymbol{x_i} \geq 0} \quad & \sum_{k \in \mathcal{K}} W_k(\boldsymbol{d_k}) - \sum_{j \in \mathcal{T}} C_j(\boldsymbol{v_j}) + \sum_{i \in \mathcal{H}} V_i(\boldsymbol{x_i}) \\ & + \pi^T \left(\sum_{i \in \mathcal{H}} U_i\left(\boldsymbol{u_i}\right) + \sum_{j \in \mathcal{T}} \boldsymbol{v_j} - \sum_{k \in \mathcal{K}} \boldsymbol{d_k} \right) \\ \text{s.t.} \quad & \boldsymbol{x_i} = \boldsymbol{x_i^0} - \boldsymbol{u_i} + \boldsymbol{h_i^1}, \quad i \in \mathcal{H} \end{aligned}$$

Problem then decouples into multiple optimizations

$$\sum_{k \in \mathcal{K}} \max_{\mathbf{d}_{k} \geq 0} \left(W_{k} \left(\mathbf{d}_{k} \right) - \pi^{T} \mathbf{d}_{k} \right) + \sum_{j \in \mathcal{T}} \max_{\mathbf{v}_{j} \geq 0} \left(\pi^{T} \mathbf{v}_{j} - C_{j}(\mathbf{v}_{j}) \right) + \sum_{i \in \mathcal{H}} \max_{\mathbf{u}_{i}, \mathbf{x}_{i} \geq 0} \left(\pi^{T} U_{i} \left(\mathbf{u}_{i} \right) + V_{i}(\mathbf{x}_{i}) \right)$$
s.t. $\mathbf{x}_{i} = \mathbf{x}_{i}^{0} - \mathbf{u}_{i} + h_{i}^{1}$

SO equivalent to CE (price takers)

Perfectly competitive (Walrasian) equilibrium is a MOPEC

CE: Consumers
$$k \in \mathcal{K}$$
 solve $\mathsf{CP}(k) : \max_{\substack{d_k \geq 0}} \ W_k\left(\frac{d_k}{d_k}\right) - \pi^T d_k$

Thermal plants $j \in \mathcal{T}$ solve $\mathsf{TP}(j) : \max_{\substack{v_j \geq 0}} \ \pi^T v_j - C_j(v_j)$

Hydro plants $i \in \mathcal{H}$ solve $\mathsf{HP}(i) : \max_{\substack{u_i, x_i \geq 0}} \ \pi^T U_i\left(u_i\right) + V_i(x_i)$

s.t. $x_i = x_i^0 - u_i + h_i^1$

$$0 \leq \pi \perp \sum_{i \in \mathcal{H}} \textit{U}_i\left(\frac{\textit{u}_i}{\textit{u}_i} \right) + \sum_{j \in \mathcal{T}} \textit{v}_j \geq \sum_{k \in \mathcal{K}} \frac{\textit{d}_k}{\textit{d}_k}.$$

But in practice there is a gap between SO and CE. How to explain?



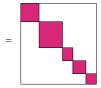
MOPEC

$$\min_{\mathbf{x}_i} \theta_i(\mathbf{x}_i, \mathbf{x}_{-i}, \pi) \text{ s.t. } g_i(\mathbf{x}_i, \mathbf{x}_{-i}, \pi) \leq 0, \forall i$$

$$\pi$$
 solves VI($h(x, \cdot), C$)

equilibrium
min theta(1) x(1) g(1)
...
min theta(m) x(m) g(m)





- (Generalized) Nash
- Reformulate optimization problem as first order conditions (complementarity)
- Use nonsmooth Newton methods to solve
- Solve overall problem using "individual optimizations"?



vi h pi cons

Perfect competition

$$\frac{\max_{x_i} \pi^T x_i - c_i(x_i)}{\text{s.t. } B_i x_i = b_i, x_i \ge 0} \qquad \frac{\text{technical constr}}{0 \le \pi \perp \sum_i x_i - d(\pi) \ge 0}$$

- When there are many agents, assume none can affect π by themselves
- Each agent is a price taker
- Two agents, $d(\pi) = 24 \pi$, $c_1 = 3$, $c_2 = 2$
- KKT(1) + KKT(2) + Market Clearing gives Complementarity**Problem**
- $x_1 = 0$, $x_2 = 22$, $\pi = 2$

Cournot: two agents (duopoly)

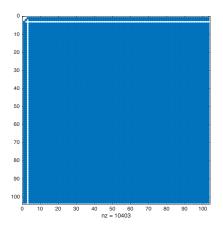
$$\max_{x_i} p(\sum_j x_j)^T x_i - c_i(x_i)$$
 profit
s.t. $B_i x_i = b_i, x_i \ge 0$ technical constr

- Cournot: assume each can affect π by choice of x_i
- Inverse demand p(q): $\pi = p(q) \iff q = d(\pi)$
- Two agents, same data
- KKT(1) + KKT(2) gives Complementarity Problem
- $x_1 = 20/3$, $x_2 = 23/3$, $\pi = 29/3$
- Exercise of market power (some price takers, some Cournot)

Computational issue: PATH

- Cournot model: $|\mathcal{A}| = 5$
- Size $n = |\mathcal{A}| * N_a$

Size (n)	Time (secs)
1,000	35.4
2,500	294.8
5,000	1024.6

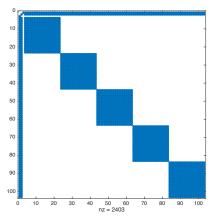


Jacobian nonzero pattern n = 100, $N_a = 20$

Computation: implicit functions

- Use implicit fn: $z(x) = \sum_{i} x_{i}$
- Generalization to F(z, x) = 0 (via adjoints)
- empinfo: implicit z F

Size (n)	Time (secs)
1,000	2.0
2,500	8.7
5,000	38.8
10,000	> 1080

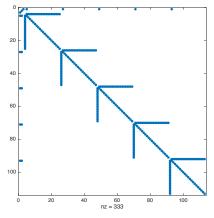


Jacobian nonzero pattern n = 100, $N_a = 20$

Computation: implicit functions and local variables

- Use implicit fn: $z(x) = \sum_{j} x_{j}$ (and local aggregation)
- Generalization to F(z, x) = 0 (via adjoints)
- empinfo: implicit z F

Size (n)	Time (secs)			
1,000	0.5			
2,500	0.8			
5,000	1.6			
10,000	3.9			
25,000	17.7			
50,000	52.3			



Jacobian nonzero pattern n = 100, $N_a = 20$

Other specializations and extensions

$$\min_{\mathbf{x}_i} \theta_i(\mathbf{x}_i, \mathbf{x}_{-i}, \pi) \text{ s.t. } g_i(\mathbf{x}_i, \mathbf{x}_{-i}, \pi) \leq 0, \forall i$$

$$\pi$$
 solves VI($h(x, \cdot), C$)

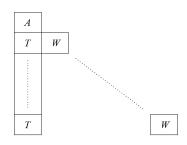
- NE: Nash equilibrium (no VI coupling constraints, $g_i(x_i)$ only)
- GNE: Generalized Nash Equilibrium (feasible sets of each players problem depends on other players variables)
- Shared constraints: some g_i 's are known to all (many) players
- Force all shared constraints to have same dual variable (VI solution)
- Can use EMP to write all these problems, and convert to MCP form
- Use models to evaluate effects of regulations and their implementation in a competitive environment

Stochastic: Agents have recourse?

- Agents face uncertainties in reservoir inflows
- Two stage stochastic programming, x^1 is here-and-now decision, recourse decisions x^2 depend on realization of a random variable
- ρ is a risk measure (e.g. expectation, CVaR)

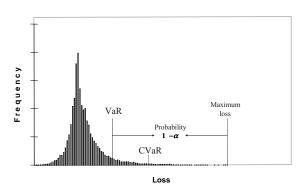
SP: min
$$c(x^1) + \rho[q^T x^2]$$

s.t. $Ax^1 = b, \quad x^1 \ge 0,$
 $T(\omega)x^1 + W(\omega)x^2(\omega) \ge d(\omega),$
 $x^2(\omega) \ge 0, \forall \omega \in \Omega.$



Risk Measures

- Modern approach to modeling risk aversion uses concept of risk measures
- \overline{CVaR}_{α} : mean of upper tail beyond α -quantile (e.g. $\alpha = 0.95$)



- mean-risk, mean deviations from quantiles, VaR, CVaR
- Much more in mathematical economics and finance literature
- Optimization approaches still valid, different objectives, varying convex/non-convex difficulty

CP:
$$\min_{d^1 \ge 0} p^1 d^1 - W(d^1)$$

TP: $\min_{v^1 \ge 0} C(v^1) - p^1 v^1$

HP: $\min_{u^1, x^1 \ge 0} - p^1 U(u^1)$

s.t. $x^1 = x^0 - u^1 + h^1$

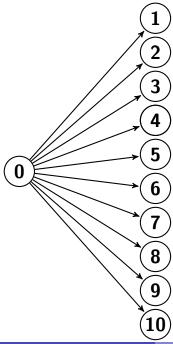
$$0 < p^1 \perp U(u^1) + v^1 > d^1$$

Two stage stochastic MOPEC (1,1,1)

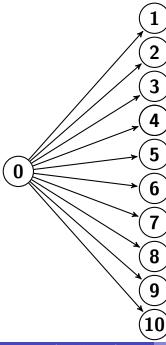
CP:
$$\min_{\substack{d^1, d_{\omega}^2 \ge 0 \\ v^1, v_{\omega}^2 \ge 0}} p^1 d^1 - W(d^1) + \rho_C \left[p_{\omega}^2 d_{\omega}^2 - W(d_{\omega}^2) \right]$$
TP: $\min_{\substack{v^1, v_{\omega}^1 \ge 0 \\ u_{\omega}^2, x_{\omega}^2 \ge 0}} C(v^1) - p^1 v^1 + \rho_T \left[C(v_{\omega}^2) - p_{\omega}^2 v^2(\omega) \right]$
HP: $\min_{\substack{u^1, x^1 \ge 0 \\ u_{\omega}^2, x_{\omega}^2 \ge 0}} - p^1 U(u^1) + \rho_H \left[-p^2(\omega) U(u_{\omega}^2) - V(x_{\omega}^2) \right]$
s.t. $x^1 = x^0 - u^1 + h^1$, $x_{\omega}^2 = x^1 - u_{\omega}^2 + h_{\omega}^2$

$$0 \le p^1 \perp U(u^1) + v^1 \ge d^1$$

$$0 \le p_{\omega}^2 \perp U(u_{\omega}^2) + v_{\omega}^2 \ge d_{\omega}^2, \forall \omega$$



- Single hydro, thermal and representative consumer
- Initial storage 10, inflow of 4 to 0, equal prob random inflows of i to node i
- Risk neutral: SO equivalent to CE (key point is that each risk set is a singleton, and that is the same as the system risk set)



- Single hydro, thermal and representative consumer
- Initial storage 10, inflow of 4 to 0, equal prob random inflows of i to node i
- Risk neutral: SO equivalent to CE
 (key point is that each risk set is a
 singleton, and that is the same as
 the system risk set)
- Each agent has its own risk measure, e.g. 0.8EV + 0.2CVaR
- Is there a system risk measure?
- Is there a system optimization problem?

$$\min \sum_{i} C(x_i^1) + \rho_i \left(C(x_i^2(\omega)) \right) ????$$

Equilibrium or optimization?

Theorem

If (d, v, u, x) solves (risk averse) SO, then there exists a probability distribution σ_k and prices p so that (d, v, u, x, p) solves (risk neutral) $CE(\sigma)$

(Observe that each agent must maximize their own expected profit using probabilities σ_k that are derived from identifying the worst outcomes as measured by SO. These will correspond to the worst outcomes for each agent only under very special circumstances)

- High initial storage level (15 units)
 - ▶ Worst case scenario is 1: lowest system cost, smallest profit for hydro
 - SO equivalent to CE
- Low initial storage level (10 units)
 - Different worst case scenarios
 - ▶ SO different to CE (for large range of demand elasticities)
- Attempt to construct agreement on what would be the worst-case outcome by trading risk

Contracts in MOPEC (Philpott/F./Wets)

- Can we modify (complete) system to have a social optimum by trading risk?
- How do we design these instruments? How many are needed? What is cost of deficiency?
- Facilitated by allowing contracts bought now, for goods delivered later (e.g. Arrow-Debreu Securities)
- Conceptually allows to transfer goods from one period to another (provides wealth retention or pricing of ancilliary services in energy market)
- Can investigate new instruments to mitigate risk, or move to system optimal solutions from equilibrium (or market) solutions

Theory and Observations

- agent problems are multistage stochastic optimization models
- perfectly competitive partial equilibrium still corresponds to a social optimum when all agents are risk neutral and share common knowledge of the probability distribution governing future inflows
- situation complicated when agents are risk averse
 - utilize stochastic process over scenario tree
 - under mild conditions a social optimum corresponds to a competitive market equilibrium if agents have time-consistent dynamic coherent risk measures and there are enough traded market instruments (over tree) to hedge inflow uncertainty
- Otherwise, must solve the stochastic equilibrium problem
- Research challenge: develop reliable algorithms for large scale decomposition approaches to MOPEC

Reserves, interruptible load, demand response

- Generators set aside capacity for "contingencies" (reserves)
- Separate energy π_d and reserve π_r prices
- Consumers may also be able to reduce consumption for short periods
- Alternative to sharp price increases during peak periods
- Constraints linking energy "bids" and reserve "bids"

$$\mathbf{v}_j + \mathbf{u}_j \leq \mathcal{U}_j, \mathbf{u}_j \leq \mathcal{B}_j \mathbf{v}_j$$

 Multiple scenarios - linking constraints on bids require "bid curve to be monotone"

Price taking: model is MOPEC

Consumption d_k , demand response r_k , energy v_j , reserves u_j , prices π

$$\begin{array}{ll} \text{Consumer} & \max_{(\boldsymbol{d}_k,r_k)\in\mathcal{C}} \text{utility}(\boldsymbol{d}_k) - \pi_{\boldsymbol{d}}{}^T\boldsymbol{d}_k + \text{profit}(\boldsymbol{r}_k,\pi_r) \\ & \text{Generator} & \max_{(\boldsymbol{v}_j,u_j)\in\mathcal{G}} \text{profit}(\boldsymbol{v}_j,\pi_{\boldsymbol{d}}) + \text{profit}(\boldsymbol{u}_j,\pi_r) \\ & \text{s.t.} & \boldsymbol{v}_j + \boldsymbol{u}_j \leq \mathcal{U}_j, \boldsymbol{u}_j \leq \mathcal{B}_j \boldsymbol{v}_j \\ & \text{Transmission} & \max_{\boldsymbol{f}\in\mathcal{F}} \text{congestion rates}(\boldsymbol{f},\pi_{\boldsymbol{d}}) \end{array}$$

Market clearing

$$0 \le \pi_d \perp \sum_{j} v_j - \sum_{k} d_k - \mathcal{A}f \ge 0$$
$$0 \le \pi_r \perp \sum_{j} u_j + \sum_{k} r_k - \mathcal{R} \ge 0$$

Large consumer is price making: MPEC

Leader/follower

Consumer max utility
$$(d_k) - \pi_d^T d_k + \operatorname{profit}(r_k, \pi_r)$$

with the constraints:

$$(d_k, r_k) \in \mathcal{C}$$
Generator $\max_{(v_j, u_j) \in \mathcal{G}'} \operatorname{profit}(v_j, \pi_d) + \operatorname{profit}(u_j, \pi_r)$
Transmission $\max_{f \in \mathcal{F}} \operatorname{congestion rates}(f, \pi_d)$
 $0 \le \pi_d \perp \sum_j v_j - \sum_k d_k - \mathcal{A}f \ge 0$
 $0 \le \pi_r \perp \sum_j u_j + \sum_k r_k - \mathcal{R} \ge 0$

Solution and observations

- Formulate as MIP, add mononticity constraints and scenarios
- New Zealand (NZEM) data, large consumer at bottom of South Island
- Expected difference percentage between "wait and see" solutions versus model solution (evaluated post optimality with simulation)

Sample Size	1	2	4	6	8
Expected diff	31.34	17.83	9.22	7.35	9.26
Standard dev	22.86	9.62	4.86	7.69	6.59
Bound gap (%)	0	0	0	12.7	24.8

- More samples better(!)
- More research to model/solve more detailed problems

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)
- dualvar (use multipliers from one agent as variables for another)
- QS functions (both in objectives and constraints)
- implicit functions and shared constraints
- Currently available within GAMS
- Some solution algorithms implemented in modeling system limitations on size, decomposition and advanced algorithms
- Can evaluate effects of regulations and their implementation in a competitive environment

Dual Representation of Risk Measures

Dual representation (of coherent r.m.) in terms of risk sets

$$\rho(Z) = \sup_{\mu \in \mathcal{D}} \mathbb{E}_{\mu}[Z]$$

- If $\mathcal{D} = \{p\}$ then $\rho(Z) = \mathbb{E}[Z]$
- If $\mathcal{D}_{\alpha,p}=\{\lambda: 0\leq \lambda_i\leq p_i/(1-lpha), \sum_i \lambda_i=1\}$, then

$$\rho(Z) = \overline{CVaR}_{\alpha}(Z)$$

Special case of a Quadratic Support Function

$$\rho(y) = \sup_{u \in U} \langle u, By + b \rangle - \frac{1}{2} \langle u, Mu \rangle$$

 EMP allows any Quadratic Support Function to be defined and facilitates a model transformation to a tractable form for solution

Addition: compose equilibria with QS functions

 Add soft penalties to objectives and/or within constraints:

$$\min_{x} \theta(x) + \rho_{O}(F(x))$$

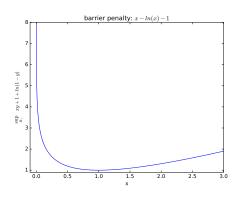
s.t. $\rho_{C}(g(x)) \le 0$

QS g rhoC udef B M

QSF cvarup F rhoO theta p

- \$batinclude QSprimal modname using emp min obj
- Allow modeler to compose QS functions automatically

- Can solve using MCP or primal reformulations
- More general conjugate functions also possible:



The link to MOPEC

$$\min_{x \in X} \theta(x) + \rho(F(x))$$

$$\rho(y) = \sup_{u \in U} \langle u, y \rangle - \frac{1}{2} \langle u, Mu \rangle$$

$$0 \in \partial \theta(x) + \nabla F(x)^{\mathsf{T}} \partial \rho(F(x)) + N_X(x)$$

$$0 \in \partial \theta(x) + \nabla F(x)^{T} u + N_{X}(x)$$

$$0 \in -u + \partial \rho(F(x)) \iff 0 \in -F(x) + Mu + N_{U}(u)$$

This is a MOPEC, and we have multiple copies of this for each agent