### Modeling and Optimization within Interacting Systems

#### Michael C. Ferris

University of Wisconsin, Madison

Joint work with Olivier Huber and Youngdae Kim

Funded by DOE-MACS Grant with Argonne National Laboratory

SAMSI, September 1, 2016

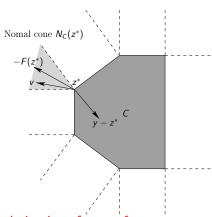
#### KKT conditions and Normal Cones

If 
$$C = \{z : g(z) \le 0\}$$
,  $g$  convex, (with CQ)

$$x^*$$
 solves  $\min_{x \in \mathcal{C}} f(x)$ 

$$\iff 0 = \nabla f(x^*) + \nabla g(x^*)\lambda, \\ 0 \le -g(x^*) \perp \lambda \ge 0$$

$$\iff 0 \in \nabla f(x^*) + N_{\mathcal{C}}(x^*)$$



Many applications where F is not the derivative of some f

# Variational Inequality (replace $\nabla f(z)$ with F(z))

- $F: \mathbb{R}^n \to \mathbb{R}^n$
- Ideally:  $C \subseteq \mathbb{R}^n$  constraint set; Often:  $C \subseteq \mathbb{R}^n$  simple bounds

$$VI(F,C): 0 \in F(z) + N_C(z)$$

- model vi / F, g /;empinfo: vi F z g
- VI generalizes many problem classes
- Nonlinear Equations: F(z) = 0 set  $C \equiv \mathbb{R}^n$
- Convex optimization:  $F(z) = \nabla f(z)$
- For MCP (rectangular VI), set  $C \equiv [I, u]^n$ .
- Can now do MPEC (as opposed to MPCC)!
- ullet Projection algorithms, robustness (evaluate F only at points in  $\mathcal C$ )

## 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

- $q = d(\pi)$ : issue is that  $\pi$  is the multiplier on the "balance" constraint
- Such multipliers (LMP's) are critical to operation of market
- ullet Can solve the problem iteratively or by writing down the KKT conditions of this QP, forming an LCP and exposing  $\pi$  to the model
- Existence, uniqueness, stability from variational analysis
- EMP does this automatically from the annotations



#### Reformulation details

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

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

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

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

#### Reformulation details

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

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

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

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

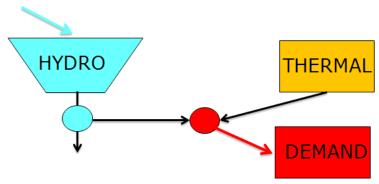
$$z = \begin{bmatrix} \pi \\ \lambda \\ x \end{bmatrix}, \quad F(z) = \begin{bmatrix} & & A \\ & B \\ -A^T & -B^T \end{bmatrix} z + \begin{bmatrix} -d(\pi) \\ -b \\ \nabla c(x) \end{bmatrix}$$

## Other applications of complementarity

- 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 VI

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

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

- $u_i$  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
- $C_j(v_j)$  denote the cost of generation by thermal plant
- $V_i(\mathbf{x}_i)$  future value of terminating with storage x (assumed separable)
- $W_k(d_k)$  utility of consumption  $d_k$

## SO equivalent to CE (price takers)

Consumers 
$$k \in \mathcal{K}$$
 solve  $\mathsf{CP}(k)$ :  $\max_{d_k \geq 0} \quad W_k\left(d_k\right) - \pi^T d_k$  Thermal plants  $j \in \mathcal{T}$  solve  $\mathsf{TP}(j)$ :  $\max_{\substack{v_j \geq 0 \\ u_i, v_i \geq 0}} \quad \pi^T v_j - C_j(v_j)$  Hydro plants  $i \in \mathcal{H}$  solve  $\mathsf{HP}(i)$ :  $\max_{\substack{u_i, v_i \geq 0 \\ u_i, v_i \geq 0}} \quad \pi^T U_i\left(u_i\right) + V_i(x_i)$  s.t.  $x_i = x_i^0 - u_i + h_i^1$ 

Perfectly competitive (Walrasian) equilibrium is a MOPEC

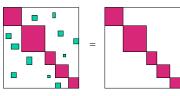
$$\begin{aligned} \mathsf{CE:} & \quad d_k \in \operatorname{arg\,max} \mathsf{CP}(k), & \quad k \in \mathcal{K}, \\ & \quad v_j \in \operatorname{arg\,max} \mathsf{TP}(j), & \quad j \in \mathcal{T}, \\ & \quad u_i, x_i \in \operatorname{arg\,max} \mathsf{HP}(i), & \quad i \in \mathcal{H}, \\ & \quad 0 \leq \pi \perp \sum_{i \in \mathcal{H}} U_i\left(u_i\right) + \sum_{j \in \mathcal{T}} v_j \geq \sum_{k \in \mathcal{K}} d_k. \end{aligned}$$

#### **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)
vi h pi cons



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



#### Perfect competition

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

- ullet When there are many agents, assume none can affect  $\pi$  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)

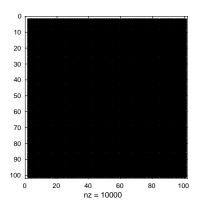
$$\max_{x_i} p(\sum_j x_j)^T x_i - c_i(x_i)$$
 profit  
s.t.  $B_i x_i = b_i, x_i \ge 0$  technical constr

- Cournot: assume each can affect p by choice of  $x_i$
- 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)
2,500	48.431
5,000	570.214

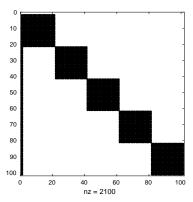


Jacobian nonzero pattern n = 100

## 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)
2,500	0.696
5,000	1.408
10,000	2.780
50,000	17.856
100,000	41.440



Jacobian nonzero pattern n = 100

### 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( $h(x, \cdot), C$ )

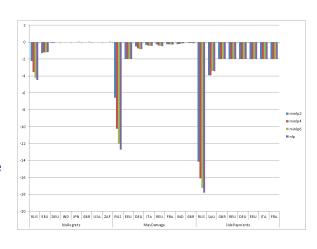
- NE: Nash equilibrium (no VI coupling constraints,  $g_i(x_i)$  only)
- GNE: Generalized Nash Equilibrium (feasible sets of each players problem depends on other players variables)
- Shared constraints: some  $g_i$ 's are known to all (many) players
- Force all shared constraints to have same dual variable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

## Optimal Sanctions (Boehringer/F./Rutherford)

- Sanctions can be modeled using similar formulations used for tariff calculations
- Model as a Nash equilibrium with players being countries (or a coalition of countries)
- Demonstrate the actual effects of different policy changes and the power of different economic instruments
- GTAP global production/trade database: 113 countries, 57 goods, 5 factors
- Coalition members strategically choose trade taxes to
  - optimize their welfare (trade war) or
  - 2 minimize Russian welfare
- Russia chooses trade taxes to maximize Russian welfare in response
- Impose (QS) constraints that limit the number of instruments used for each country

## Optimal Sanctions: Results

- Resulting Nash equilibrium with trade war, maximize damage, side payments - all have big impact on Russia
- Restricting instruments can change effects (these are the different colored bars)
- Collective (coalition) action significantly better

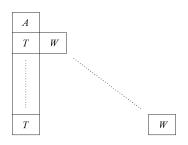


Same model can used to determine effects of Russian trade sanctions on Turkey

## Stochastic: Agents have recourse?

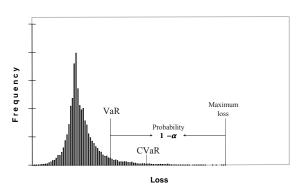
- Agents face uncertainties in reservoir inflows
- Two stage stochastic programming,  $x^1$  is here-and-now decision, recourse decisions  $x^2$  depend on realization of a random variable
- $\rho$  is a risk measure (e.g. expectation, CVaR)

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
- $\overline{CVaR}_{\alpha}$ : mean of upper tail beyond  $\alpha$ -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

#### 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  $\mathcal{D}_{\alpha,p}=\{\lambda: 0\leq \lambda_i\leq p_i/(1-\alpha), \sum_i \lambda_i=1\}$ , then

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

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:

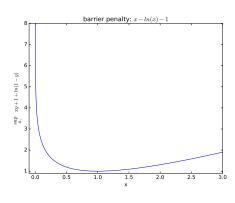
$$\min_{x} \theta(x) + \rho_O(F(x))$$
  
s.t.  $\rho_C(g(x)) \le 0$ 

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

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

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

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

$$0 \in \partial \theta(x) + \nabla F(x)^{\mathsf{T}} u + \mathsf{N}_{\mathsf{X}}(x)$$
  
$$0 \in -u + \partial \rho(F(x)) \iff 0 \in -F(x) + \mathsf{M}u + \mathsf{N}_{\mathsf{U}}(u)$$

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

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

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

# Two stage stochastic MOPEC (1,1,1)

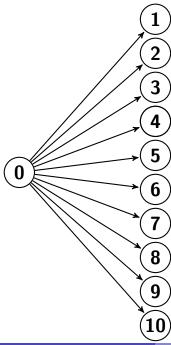
$$\begin{split} \text{CP:} & \min_{\substack{d^1,d_\omega^2 \geq 0 \\ v^1,v_\omega^2 \geq 0}} & p^1 d^1 - W(d^1) + \rho_C \left[ p_\omega^2 d_\omega^2 - W(d_\omega^2) \right] \\ \text{TP:} & \min_{\substack{v^1,v_\omega^2 \geq 0 \\ u_\omega^2,x_\omega^2 \geq 0}} & C(v^1) - p^1 v^1 + \rho_T \left[ C(v_\omega^2)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

$$0 \leq p^1 \perp U(u^1) + v^1 \geq d^1$$

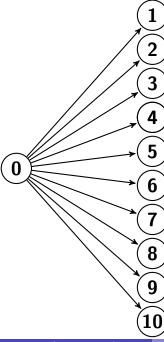
# Two stage stochastic MOPEC (1,1,1)

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

$$0 \le p^{1} \perp U(u^{1}) + v^{1} \ge d^{1}$$
$$0 \le p_{\omega}^{2} \perp U(u_{\omega}^{2}) + v_{\omega}^{2} \ge d_{\omega}^{2}, \forall \omega$$



- 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  $\sigma_k$  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  $\sigma_k$  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

#### 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