Modeling and Solving Nontraditional Optimization Problems Session 2b: Complementarity Conditions

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Chiang Mai University International Conference *Workshop* Chiang Mai, Thailand — 4-5 January 2011

Session 2b: Complementarity

Focus

 A condition closely associated with optimization that says two quantities cannot both be positive

Topics

- Motivating examples
- ✤ AMPL syntax
- More examples
- Solver strategies

Complementarity Classical Linear Program

Minimum-cost production

Complementarity Classical Linear Complementarity

Economic equilibrium

```
set PROD; # products
set ACT:
           # activities
param cost {ACT} > 0;  # cost per unit of each activity
param demand {PROD} >= 0; # units of demand for each product
param io {PROD,ACT} >= 0; # units of each product from
                           # 1 unit of each activity
var Price {PROD};
var Level {ACT};
subject to Pri_Compl {i in PROD}:
  Price[i] >= 0 complements
      sum {j in ACT} io[i,j] * Level[j] >= demand[i];
subject to Lev_Compl {j in ACT}:
  Level[j] >= 0 complements
      sum {i in PROD} Price[i] * io[i,j] <= cost[j];</pre>
```

... complementary slackness conditions for the classical linear program

Complementarity Bounded-Variable Linear Program

Minimum-cost production

```
set PROD; # products
           # activities
set ACT:
param cost {ACT} > 0;  # cost per unit of each activity
param demand {PROD} >= 0; # units of demand for each product
param io {PROD,ACT} >= 0; # units of each product from
                           # 1 unit of each activity
param level_min {ACT} > 0; # min allowed level for each activity
param level_max {ACT} > 0; # max allowed level for each activity
var Level {ACT};
minimize TotalCost:
   sum {j in ACT} cost[j] * Level[j];
subject to LevelLimits {j in ACT}:
   level_min[j] <= Level[j] <= level_max[j]</pre>
subject to Demands {i in PROD}:
   sum {j in ACT} io[i,j] * Level[j] >= demand[i];
```

Complementarity Mixed Linear Complementarity

Economic equilibrium with bounded variables

```
set PROD; # products
           # activities
set ACT:
param cost {ACT} > 0;  # cost per unit
param demand {PROD} >= 0; # units of demand
param io {PROD,ACT} >= 0; # units of product per unit of activity
param level_min {ACT} > 0; # min allowed level for each activity
param level_max {ACT} > 0; # max allowed level for each activity
var Price {PROD};
var Level {ACT};
subject to Pri_Compl {i in PROD}:
  Price[i] >= 0 complements
      sum {j in ACT} io[i,j] * Level[j] >= demand[i];
subject to Lev_Compl {j in ACT}:
   level_min[j] <= Level[j] <= level_max[j] complements</pre>
      cost[j] - sum {i in PROD} Price[i] * io[i,j];
```

... generalized complementary slackness conditions

Complementarity Nonlinear Complementarity

Equilibrium with price-dependent demands

```
set PROD; # products
set ACT; # activities
param cost {ACT} > 0;  # cost per unit
param demand {PROD} >= 0; # units of demand
param io {PROD,ACT} >= 0; # units of product per unit of activity
param demzero {PROD} > 0; # intercept and slope of the demand
param demrate {PROD} >= 0; # as a function of price
var Price {PROD};
var Level {ACT};
subject to Pri_Compl {i in PROD}:
  Price[i] >= 0 complements
      sum {j in ACT} io[i,j] * Level[j]
         >= demzero[i] + demrate[i] * Price[i];
subject to Lev_Compl {j in ACT}:
  Level[j] >= 0 complements
      sum {i in PROD} Price[i] * io[i,j] <= cost[j];</pre>
```

... not equivalent to a linear program

Complementarity AMPL's complements operator

Two single inequalities

single-ineq1 complements single-ineq2

Both inequalities must hold, at least one at equality

One double inequality

double-ineq complements expr expr complements double-ineq

The double-inequality must hold, and

if at lower limit then $expr \ge 0$, if at upper limit then $expr \le 0$, if between limits then expr = 0

One equality

equality complements expr expr complements equality

The equality must hold (included for completeness)

Complementarity Logic vs. New Operator

Using complements operator

```
subject to delct {cr in creg, u in users}:
```

0 <= ct[cr,u] complements</pre>

```
ctcost[cr,u] + cv[cr] >= p["C",u];
```

Using logic operators

```
subject to delct {cr in creg, u in users}:
0 <= ct[cr,u] and
ctcost[cr,u] + cv[cr] >= p["C",u] and
(0 = ct[cr,u] or ctcost[cr,u] + cv[cr] = p["C",u]);
```

```
... it's a tradeoff, of course
```

Complementarity **Examples** (CONT'd)

```
Prices of coal shipments
```

subject to delct {cr in creg, u in users}:

0 <= ct[cr,u] complements</pre>

ctcost[cr,u] + cv[cr] >= p["C",u];

Height of membrane

```
subject to dv {i in 1..M, j in 1..N}:
    lb[i,j] <= v[i,j] <= ub[i,j] complements
        (dy/dx) * (2*v[i,j] - v[i+1,j] - v[i-1,j])
      + (dx/dy) * (2*v[i,j] - v[i,j+1] - v[i,j-1])
      - c * dx * dy ;</pre>
```

... more at Complementarity Problem Net http://www.cs.wisc.edu/cpnet/

Optimization Examples (MPECs)

Cournot Nash equilibrium (gnash1m)

g1: 0 <= y[1] <= L	complements	1[1];
g3: 0 <= y[2] <= L	complements	1[2];
g5: 0 <= y[3] <= L	complements	1[3];
g7: 0 <= y[4] <= L	complements	1[4];

Min area packaging membrane (pac-comp1)

```
obst {i in int_nodes}:
    0 <= s1[i]
        complements
        u[i] - xi[i] - c*(l[i] - Au[i]) >= 0;
```

... more at Mac MPEC www-unix.mcs.anl.gov/~leyffer/MacMPEC/

Solving

"Square" systems

* # of variables =
 # of complementarity constraints +
 # of equality constraints

Transformation to a simpler canonical form required

MPECs (or MPCCs)

- Mathematical programs with equilibrium constraints (or . . . with complementarity constraints)
- No restriction on numbers of variables & constraints
- Objective functions permitted

... solvers continuing to emerge

Solving MPECs

Square systems: well understood * PATH is a well-known solver

Theory of MPECs: terrible

- Easily convert to smooth functions, but . . .
- Constraint qualifications are violated at every feasible point

Practice with MPECs: promising

- * *Convert* to an equivalent smooth problem
- Apply a standard method for nonlinearly constrained nonlinear optimization
 - ... choice of conversion depends on type of solver

Solving Representative MPEC Conversions

$$0 \le z_1 \perp z_2 \ge 0$$

$$z_1 \ge 0$$

$$z_2 \ge 0$$

$$z_1^T z_2 \le 0$$

$$\sqrt{(z_{1i} - z_{2i})^2 + 4\mu} - z_{1i} - z_{2i} = 0$$

$$\sqrt{z_{1i}^2 + z_{2i}^2 + \mu} - z_{1i} - z_{2i} = 0$$

Solving Representative MPEC Conversions

 $\begin{array}{ll} \text{minimize} & f(x)\\ \text{subject to} & c_i(x) = 0, \quad i \in \mathcal{E}\\ & c_i(x) \ge 0, \quad i \in \mathcal{I}\\ & 0 \le x_1 \perp x_2 \ge 0. \end{array}$

minimize	f(x)
subject to	$c_i(x)=0, i\in \mathcal{E}$
	$c_i(x) \geq 0, i \in \mathcal{I}$
	$x_1 \ge 0, x_2 \ge 0$
	$x_{1i}x_{2i} \le 0 i = 1, \dots, p.$

Alternative penalty formulation



Solving Carrying Out the MPEC Conversions

Same theme as before

- * **Detection:** Look for complements operators in constraints
- * Transformation: Convert each such constraint
 - * different for each solver
- * **Implementation:** Recursive tree walks . . .