

# KNITRO 5.1 for Nonlinear Optimization

Todd Plantenga, Richard Waltz  
Ziena Optimization, Inc.

May, 2007

# Outline

---

- KNITRO overview
  - Product, where to use it
  - Three solvers
- Release history
  - KNITRO 5.0 enhancements
  - KNITRO 5.1 enhancements
- Examples
  - Strategic bidding with MPCCs
  - Sample AMPL code

# KNITRO Overview

# KNITRO Product

---

- Solve mathematical optimization problems
- **Robust** (3 different solvers)
- **Large-scale** (sparse linear algebra)
- **Embeddable** (multiple platforms, thread-safe)
- **Commercially supported** (Ziena since 2001)

# Mathematical Problem Form

---

- KNITRO solves problems that can be written as

$$\begin{aligned} & \min_{x \in \mathbf{R}^n} && f(x) \\ & \text{subject to} && g_i(x) \geq 0 \quad i \in I \\ & && h_j(x) = 0 \quad j \in E \\ & && l_i \leq x_i \leq u_i \end{aligned}$$

- $x$  must be continuous (no integer variables)
- $f(x)$ ,  $g(x)$ ,  $h(x)$  must be smooth
- No convexity requirement

# Optimization Problem Types

---

- General nonlinear problems with constraints
- Unconstrained
- Bound constrained
- Equality constrained
- Heuristics for better performance on
  - Complementarity constraints (MPCCs)
  - Nonlinear systems of equations
  - Least squares (linear and nonlinear)
  - Linear programs (LPs)
  - Quadratic programs (convex and nonconvex)

# Where is KNITRO Used?

---

Optimal power flow

Strategic bidding, option pricing

Portfolio optimization, financial investments

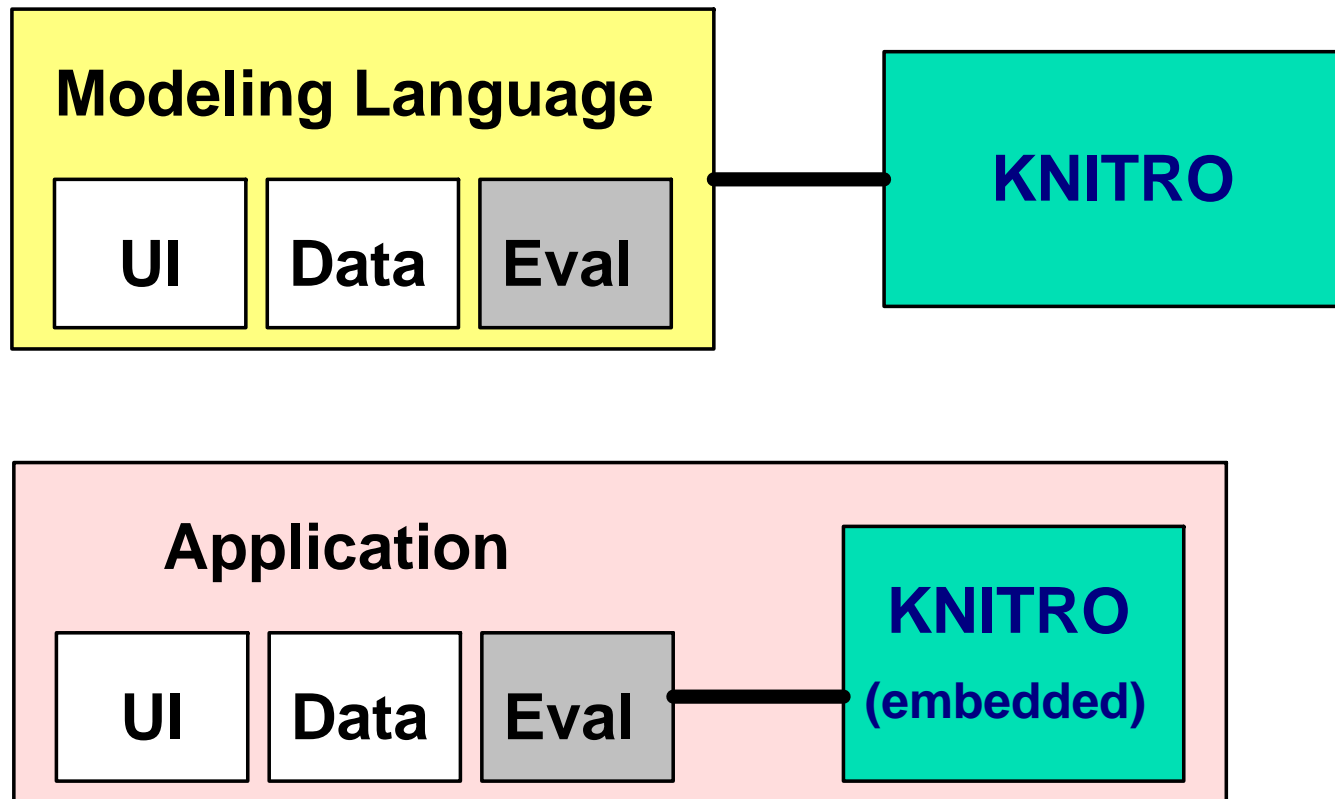
Semiconductor circuit layout

Chemical process control

Engineering design

Academic research (many applications)

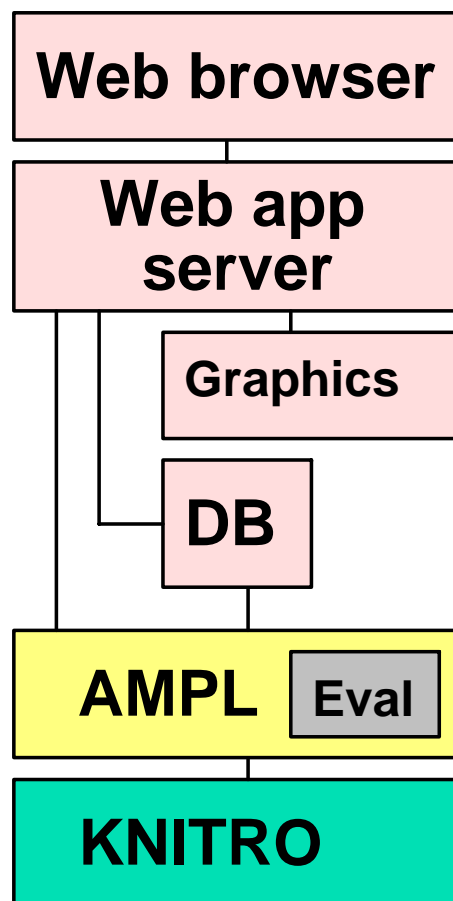
# How is KNITRO Used?



“Eval” computes partial derivative values

# Dynamic Application Example

Users modify problem, KNITRO solves in real time



## Example tech stack:

- Browser with Javascript
- Apache/JBoss or Apache/Tomcat Java-based application server
- Graphics tool to generate plot images
- Data from RDBMS
- AMPL for modeling
- KNITRO for solving

# Why Buy KNITRO?

---

- Power of 3 different solvers
- Large-scale performance
- Commercial support
- Ongoing state-of-the-art development

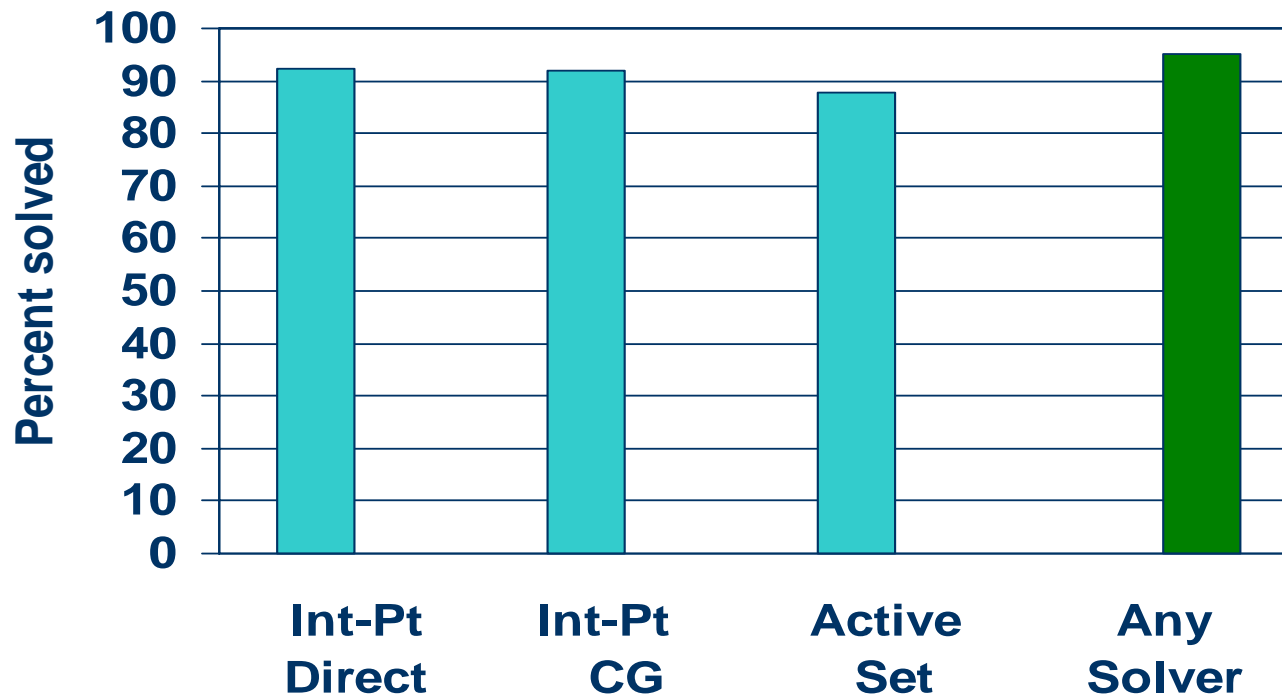
## 3 Different Solvers/Algorithms

---

- Interior-point / barrier
  - **KNITRO Interior/Direct** (alg=1, since release 2.0)  
(handles ill-conditioned problems)
  - **KNITRO Interior/CG** (alg=2, since release 1.0)  
(handles large/dense Hessians)
- Active Set SLQP
  - **KNITRO Active Set** (alg=3, since release 4.0)  
(able to warm start, crossover from Interior-point solvers)
- All state-of-the-art. KNITRO's underlying theory, published in academic papers by Ziena's founders, is cited in over 200 other publications (Google Scholar).

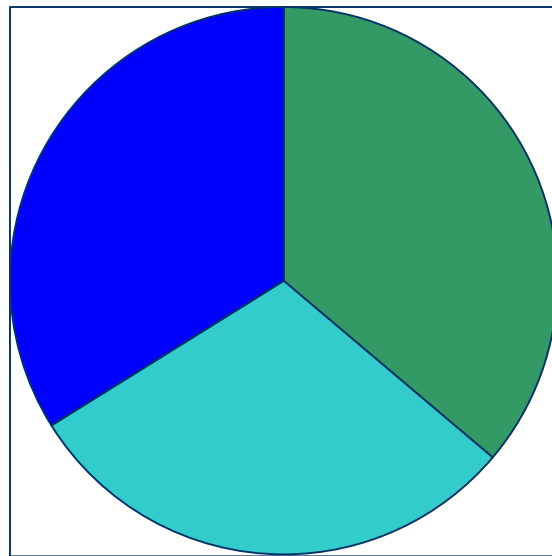
## 3 Solvers are more Robust

- 920 test problems
- 95.2% can be solved by at least one solver

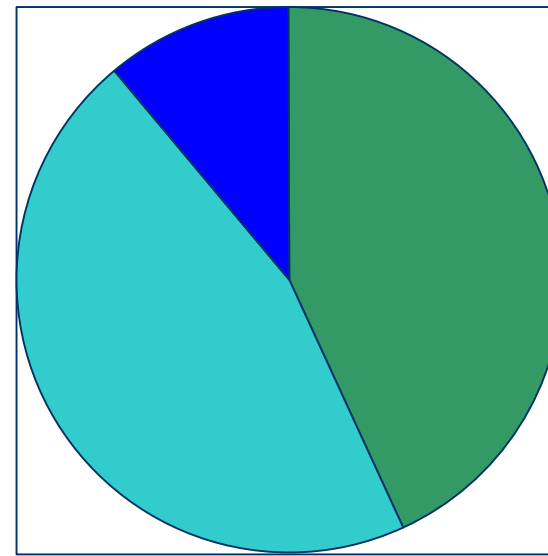


## 3 Solvers are more Efficient

- 920 test problems
- “Best” solver when both find a solution:



Function Evaluations



CPU Time



# When to use Interior/Direct

- Second derivatives (Hessian) available
- Often the fastest
- Better than Interior/CG if ill-conditioned:

Problem BQPGauss (2003 variables)

Solver	iters	time	time/iter
<b>Interior/Direct</b>	19	3	0.2
Interior/CG	27	1310	48.5

# When to use Interior/CG

---

- Second derivatives (Hessian) not available
- Hessian available but large (Interior/Direct spends too much time factorizing):

Problem CVXQP2 (10,000 variables)

Solver	iters	time	time/iter
Interior/Direct	14	2638	188.4
<b>Interior/CG</b>	11	401	36.5

# When to use Active Set

- Warm start: a good start point is known, or a series of similar problems are solved

Example: solve for different values of a constraint parameter.  
 Active Set takes fewer iterations (problem HS66).

Parameter	Interior/Direct	Active Set
0.0 (cold start)	115	105
0.4 (warm)	13	3
0.8 (warm)	11	3
0.0 (warm)	10	1

# KNITRO is Flexible Software

---

- Full platform support for
  - Windows: 32-bit and 64-bit
  - Linux: 32-bit and 64-bit
  - Mac OS X: x86 and PowerPC
  - Solaris: SPARC
- Simple API for applications coded in
  - C, C++, Fortran, Java
- Available as a solver for
  - AMPL
  - AIMMS
  - Frontline (Excel spreadsheets)
  - GAMS
  - Mathematica
  - TOMLAB (MATLAB and LabView)

# KNITRO is Commercial Software

---

- Production libraries
  - Thread-safe for embedding
  - Debugging capabilities
  - Examples provided in C, C++, Fortran, Java
- Ziena incorporated in 2001
- Ziena license manager
- Ziena support (Annual Maintenance Service)
  - Software assistance, bug fixes
  - Optimization problem formulation
  - Free upgrades for new features

# KNITRO Availability

---

- <http://www.ziena.com>
- **Free** “student” download
  - 300 variable limit
- **Free** 30-day trial of full strength version
- Academic pricing for University research

# KNITRO Release History

# KNITRO Release History

---

## 5.1 – Currently Available

5.1.0 (Nov 2006) – minor release

## 5.0 – No longer available (but still supported)

5.0.3 (Jun 2006) – bug fix

5.0.0 (Feb 2006) – major release

## 4.0 – No longer available (but still supported)

4.0.0 (Oct 2004) – major release

## 3.1 – No longer available

3.1.0 (Sep 2003) – minor release

3.0.0 (Apr 2003) – major release

(see [http://www.ziena.com/release\\_hist.htm](http://www.ziena.com/release_hist.htm))

# KNITRO 5.0 Enhancements

## New in KNITRO 5.0

---

- Capability to solve **MPCC** problems
- **Crossover** from an interior-point solution to the Active Set solver for highly accurate solutions
- **Multi-start** option for trying to find the global solution
- Additional barrier update rule
- Streamlined programming interface
- Built-in derivative checking
- Improvements in algorithm efficiency and robustness

# MPCC Problems

**Math Programming with Complementarity Constraints**  
 aka: Equilibrium Constraints / MPEC / NCP

$$\begin{aligned} & \min f(x) \\ & \text{s.t. } g(x) \geq 0 \\ & \quad h(x) = 0 \end{aligned}$$

$$0 \leq x_1 \perp x_2 \geq 0$$

Applications:

- Strategic bidding
- Economic models
- Contact problems
- Traffic equilibrium
- Disjunctive conditions

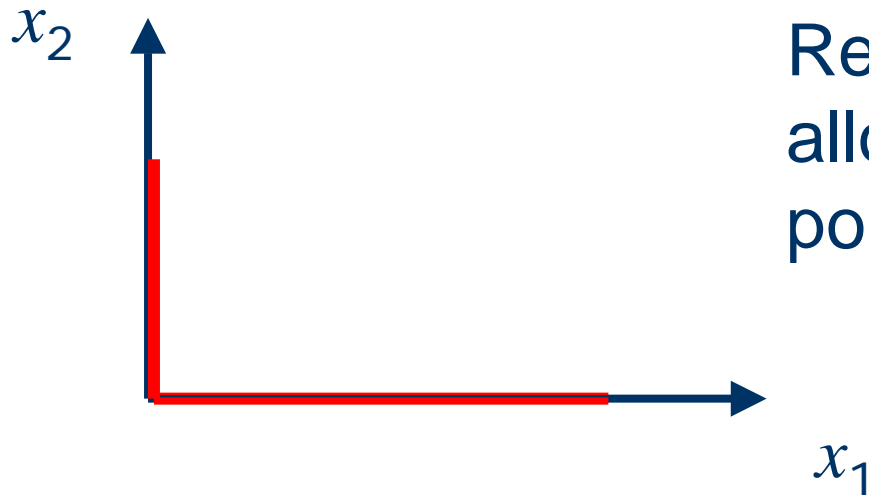
The complementarity condition means

$$\begin{aligned} & x_1 \geq 0, x_2 \geq 0 \quad \text{and} \\ & \text{either } x_1 = 0 \quad \text{or } x_2 = 0 \quad (\text{equivalently, } x_1 x_2 = 0) \end{aligned}$$

# Why is MPCC Hard?

$$\begin{aligned} &\min f(x) \\ &\text{s.t. } g(x) \geq 0 \\ &h(x) = 0 \end{aligned}$$

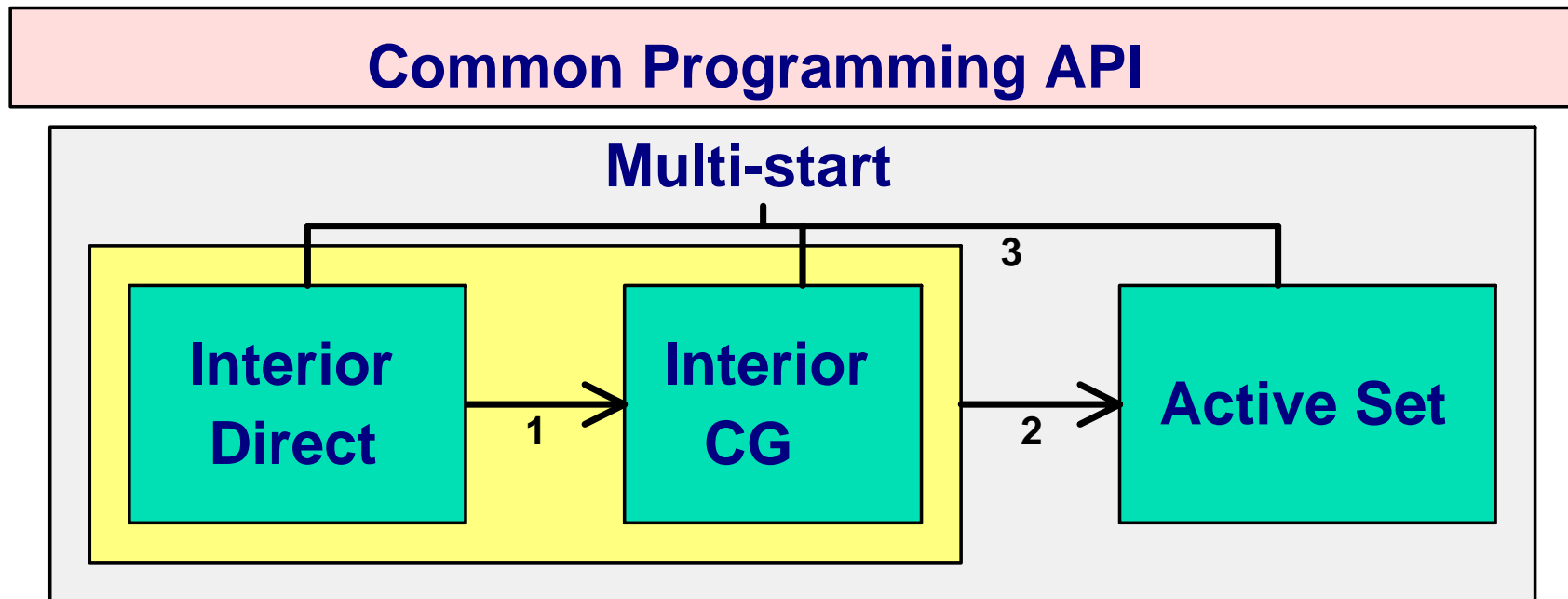
$$0 \leq x_1 \perp x_2 \geq 0$$



- Constraint qualification fails at feasible points
- Multipliers undefined, no interior path

Recent research breakthroughs allow reformulation for interior-point solvers

# Solver Interactions



1. Interior/Direct falls back on Interior/CG for improved robustness
2. Interior solvers can cross over to Active set
3. Multi-start searches for best local minima

# Crossover

---

Crossover cleans up the Interior-point solution by switching to the Active Set solver at the end

- Cleanly identifies active set of constraints
- More accurate solution and sensitivities
- Standard for LP but very difficult for NLP because it requires **2 solvers**
- Can be switched off as there may be overhead costs

# Crossover Example (HS35)

Interior-point solution (6 interior-point iterations):

Constraint Vector		Lagrange Multipliers	
$c[0] = 2.99999960166e+000,$		$\text{lambda}[0] = 2.22222343515e-001$	
Solution Vector			
$x[0] = 1.33333347124e+000,$		$\text{lambda}[1] = -3.43504445156e-008$	
$x[1] = 7.77777685286e-001,$		$\text{lambda}[2] = -1.71771265657e-008$	
$x[2] = 4.44444222565e-001,$		$\text{lambda}[3] = -3.83389961963e-008$	

Crossover solution (1 additional crossover iteration):

Constraint Vector		Lagrange Multipliers	
$c[0] = 3.00000000000e+000,$		$\text{lambda}[0] = 2.22222306741e-001$	
Solution Vector			
$x[0] = 1.33333333333e+000,$		$\text{lambda}[1] = 0.00000000000e+000$	
$x[1] = 7.77777777778e-001,$		$\text{lambda}[2] = 0.00000000000e+000$	
$x[2] = 4.44444444444e-001,$		$\text{lambda}[3] = 0.00000000000e+000$	

# Multi-start

---

- KNITRO solvers converge to a local minimum
- Different start points may go to different minima
- Efficient restarting (especially Active Set solver)
- Framework in place for sophisticated global search

# KNITRO 5.1 Enhancements

## New in KNITRO 5.1

---

- New options to improve Interior-point performance
- **Intel MKL** (Math Kernel Library) included for 20-30% speedup on BLAS/LAPACK operations
- **Newpoint** capability saves iterates, especially useful when progress is slow
- **Multi-start** generation of new start points improved with several new options
- **Mathematica** interface provides compiled object alternative for large-scale problems

# Examples

# Strategic Bidding for Electricity

---

- Your company sells electric power
- You and other producers submit competitive bids to generate power
- An ISO purchases at a single “spot price”
- Your strategic guidance:
  - submit low bids  $\leq$  spot price
  - submit high bids to drive up the spot price

# Strategic Bidding for Electricity

$b_{i,j}$  bid price by company  $i$ , plant  $j$

$c_{i,j}$  generating cost per kilowatt

$G_{i,j}$  generating capacity

$g_{i,j}$  kilowatts to generate for ISO

ISO problem:

$$\begin{aligned} & \min_{g_{i,j}} && \sum_{i,j} b_{i,j} g_{i,j} \\ & \text{subject to} && 0 \leq g_{i,j} \leq G_{i,j} \\ & && \sum_{i,j} g_{i,j} \geq \text{Demand} \end{aligned}$$

Spot price  $\pi_D$  is multiplier

# Strategic Bidding for Electricity

---

Your problem ( $i = 1$ ):

$$\begin{aligned} & \max_{b_j} \quad \sum_j (\pi_D - c_j) g_j \\ & \text{subject to} \quad \begin{cases} c_j \leq b_j \\ \text{ISO solution} \end{cases} \end{aligned}$$

ISO problem:  $b_{i,j} \rightarrow g_{i,j}, \pi_D$

- Bilevel linear optimization problem
- Can solve iteratively, but converges slowly

# Strategic Bidding for Electricity

- Replace ISO problem with KKT conditions
- Solve as a nonlinear MPEC with KNITRO 5.0

$$\begin{aligned}
 b_{i,j} - \pi_D + \pi_{i,j}^U - \pi_{i,j}^L &= 0 \quad \forall i, j \\
 0 \leq g_{i,j} \leq G_{i,j} &\quad \forall i, j \\
 \sum_{i,j} g_{i,j} &\geq Demand
 \end{aligned}$$

$$0 \leq \pi_D \perp \sum_{i,j} g_{i,j} - Demand \geq 0$$

$$0 \leq \pi^U \perp G - g \geq 0$$

$$0 \leq \pi^L \perp g \geq 0$$

# Strategic Bidding for Electricity

---

- What about bids from competitors?
- Use stochastic optimization
  - Same nonlinear MPEC formulation
  - Different scenarios for bids
  - Larger, but KNITRO handles large-scale problems

# AMPL Interface

---

- Algebraic Modeling and Programming Language
- Intuitive syntax
- Interactive user interface
- Extremely efficient evaluation of partial derivatives for optimization solvers
- File I/O, database connectivity
- Lightweight software component
- Sold by Ziena, same license manager as KNITRO

# AMPL Interface Sample Code

---

```
# Example nonlinear AMPL model.

# Define variables and make them non-negative.
var x{j in 1..3} >= 0;

# Define the objective function to be minimized.
minimize obj:
    1000 - x[1]^2 - 2*x[2]^2 - x[3]^2 - x[1]*x[2] - x[1]*x[3];

# Equality constraint.
s.t.  c1: 8*x[1] + 14*x[2] + 7*x[3] - 56 = 0;

# Inequality constraint.
s.t.  c2: x[1]^2 + x[2]^2 + x[3]^2 - 25 >= 0;

# Define initial point.
data;
let x[1] := 2;
let x[2] := 2;
let x[3] := 2;
```

# AMPL Interface User Session

---

```
AMPL> option solver knitroampl;  
AMPL> option knitro_options "outlev=0 alg=2";  
AMPL> model testproblem.mod;  
AMPL> solve;
```

```
KNITRO 5.1.0: LOCALLY OPTIMAL SOLUTION FOUND.  
objective 9.360004e+02; feasibility error 0.000000e+00  
6 major iterations; 7 function evaluations
```

```
AMPL> display x[3];  
x[3] = 7.99997  
AMPL> option knitro_options "outlev=0 alg=2 maxcrossit=1";  
AMPL> solve;
```

```
KNITRO 5.1.0: LOCALLY OPTIMAL SOLUTION FOUND.  
objective 9.360000e+02; feasibility error 0.000000e+00  
5 major iterations; 6 function evaluations
```

```
AMPL> display x[3];  
x[3] = 8
```