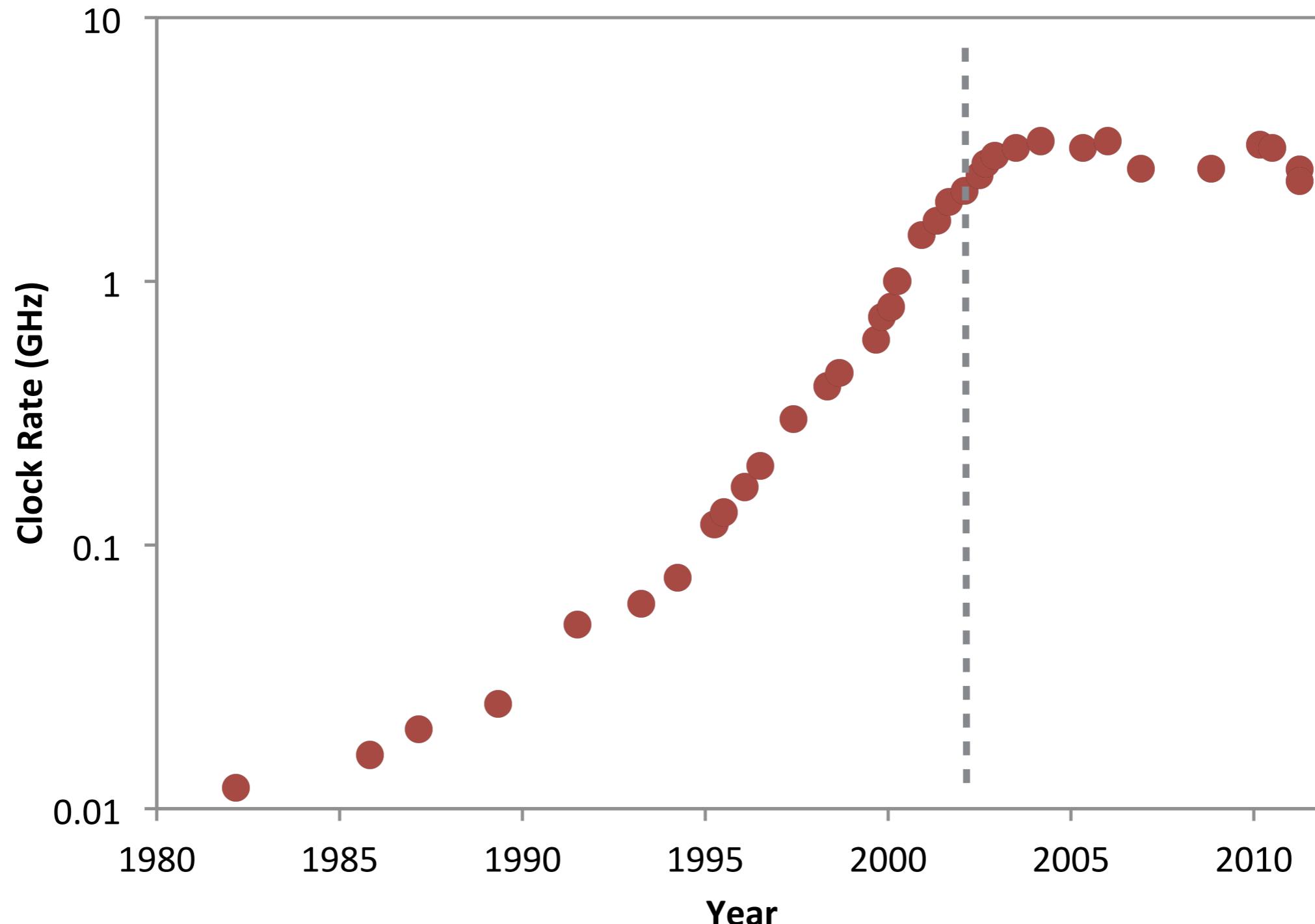


Parallel Solution of a Nonlinear Stochastic Programming Formulation for the N-1 Contingency-Constrained ACOPF Problem

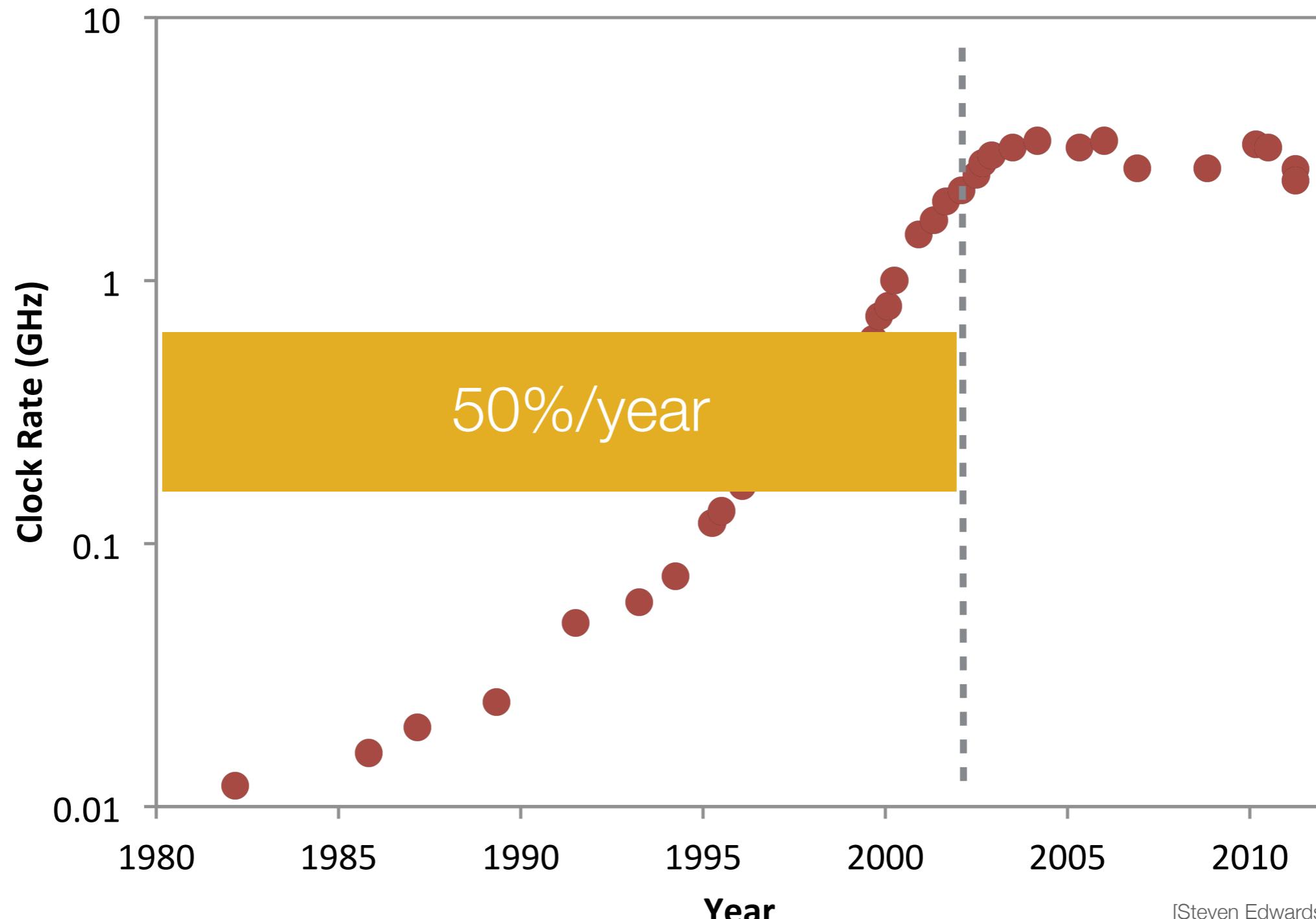
Carl D. Laird, Associate Professor, Purdue University
Anya Castillo, Cesar Silva-Monroy, Jean-Paul Watson

Laird Research Group: <http://allthingsoptimal.com>

Landscape of Scientific Computing

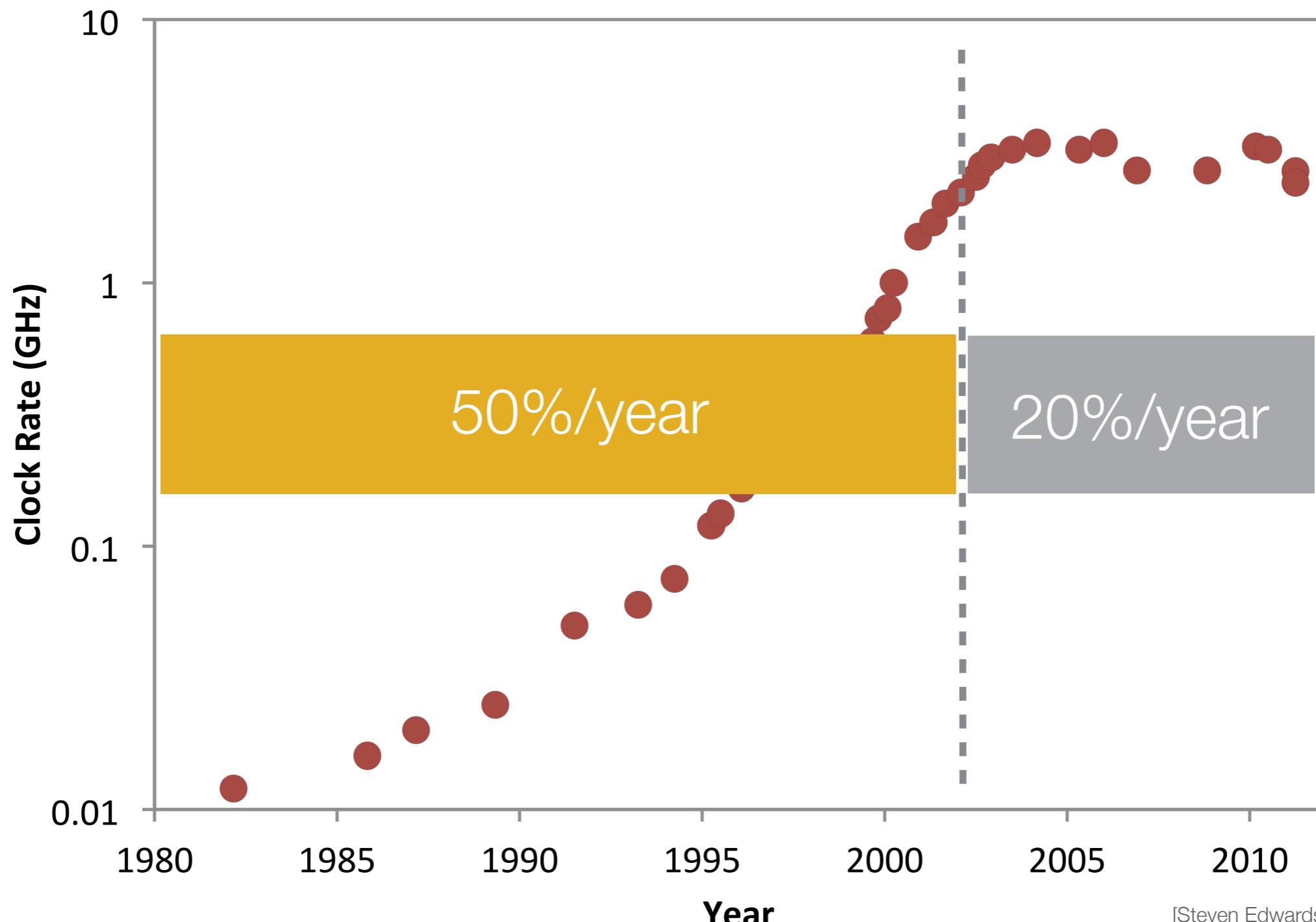


Landscape of Scientific Computing



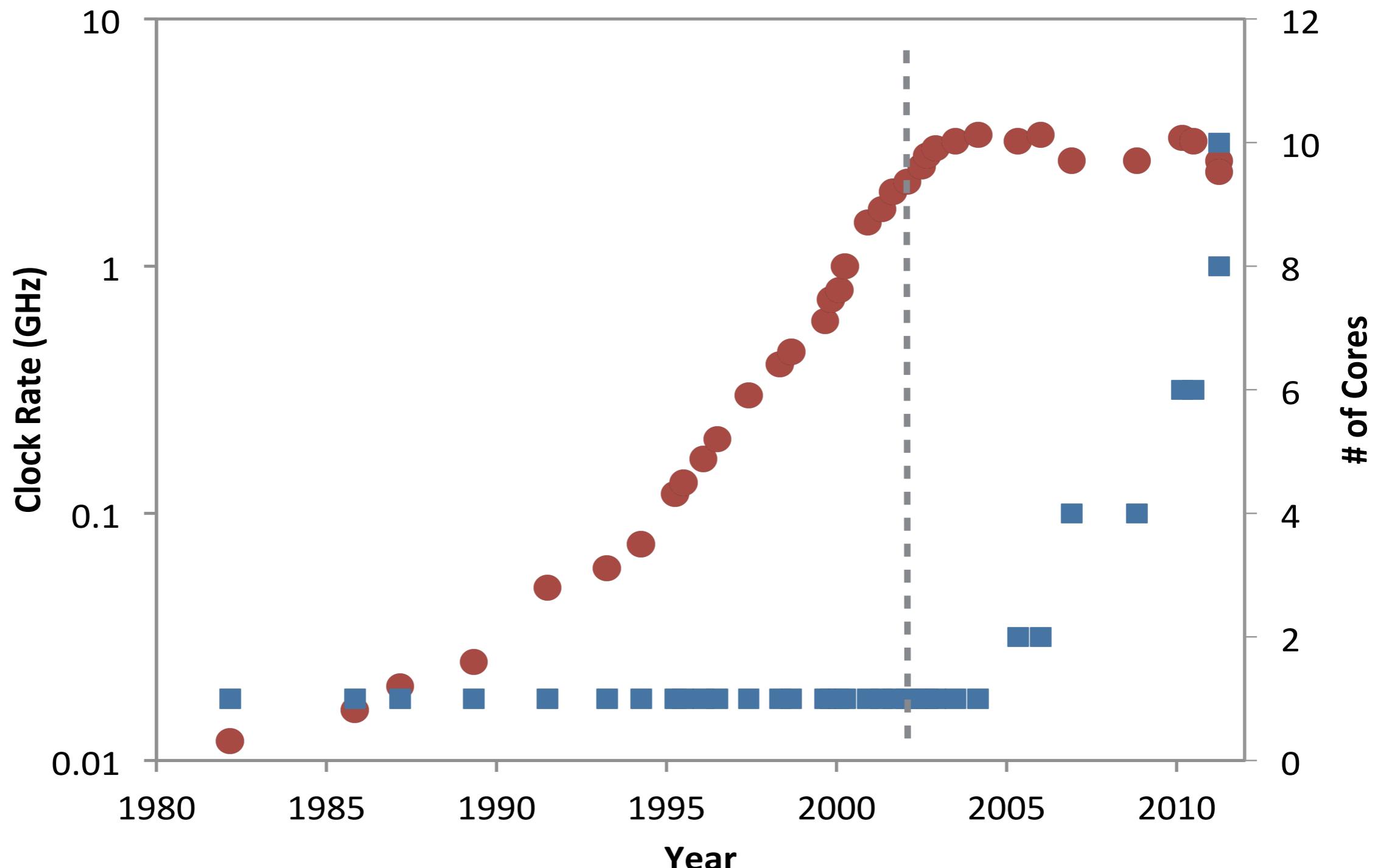
[Steven Edwards, Columbia University]

Landscape of Scientific Computing



[Steven Edwards, Columbia University]

Landscape of Scientific Computing



Landscape of Scientific Computing



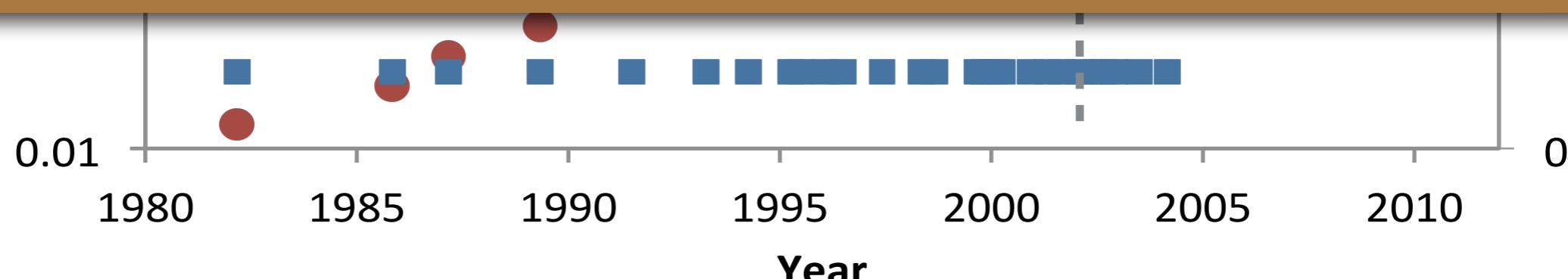
10

12

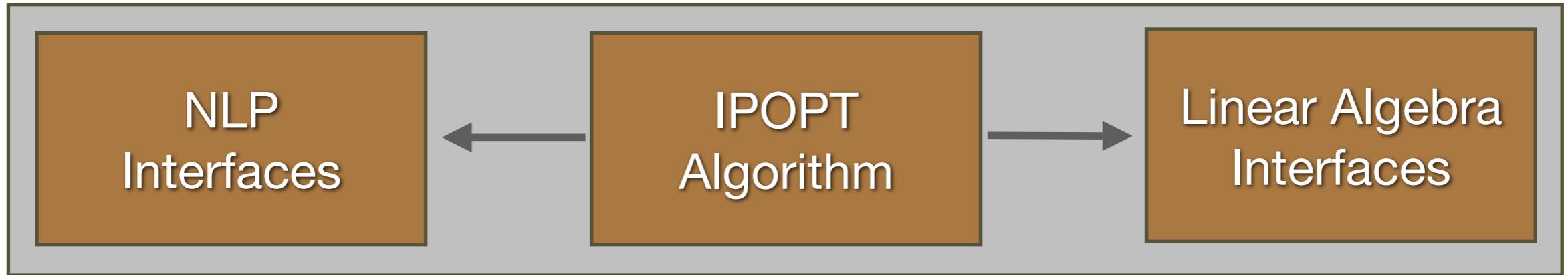
Clock-rate, one source of hardware speed improvement has stagnated.

Hardware manufacturers are shifting focus to energy efficiency and parallel architectures.

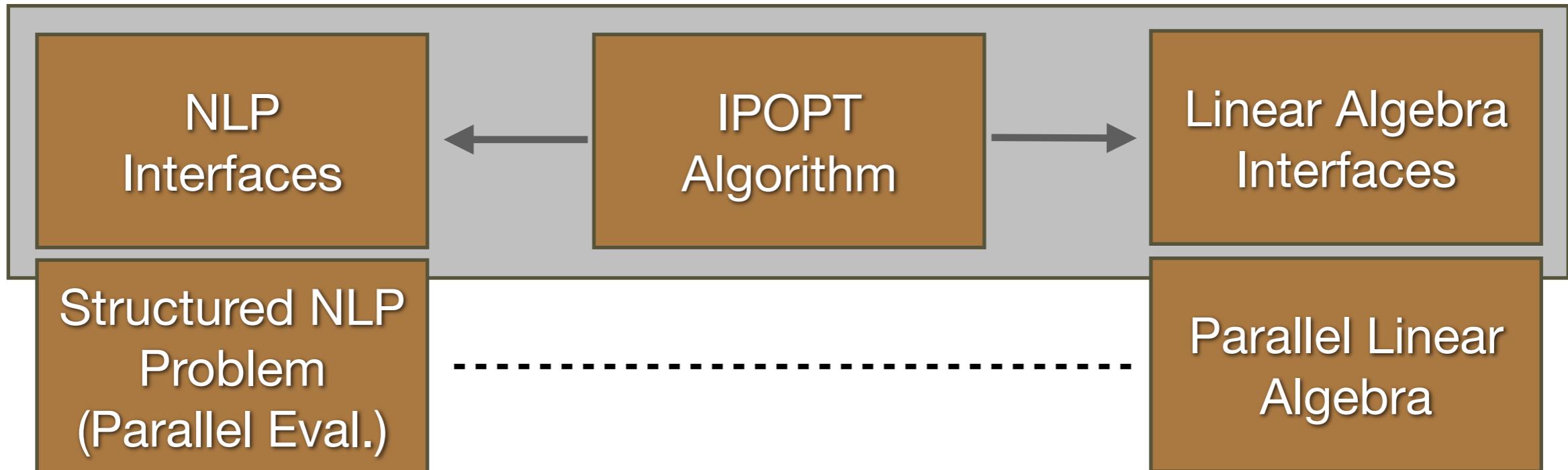
Performance improvement requires parallel algorithms (multi-core PCs, clusters, and emerging architectures).



Parallel Interior-Point Methods



Parallel Interior-Point Methods

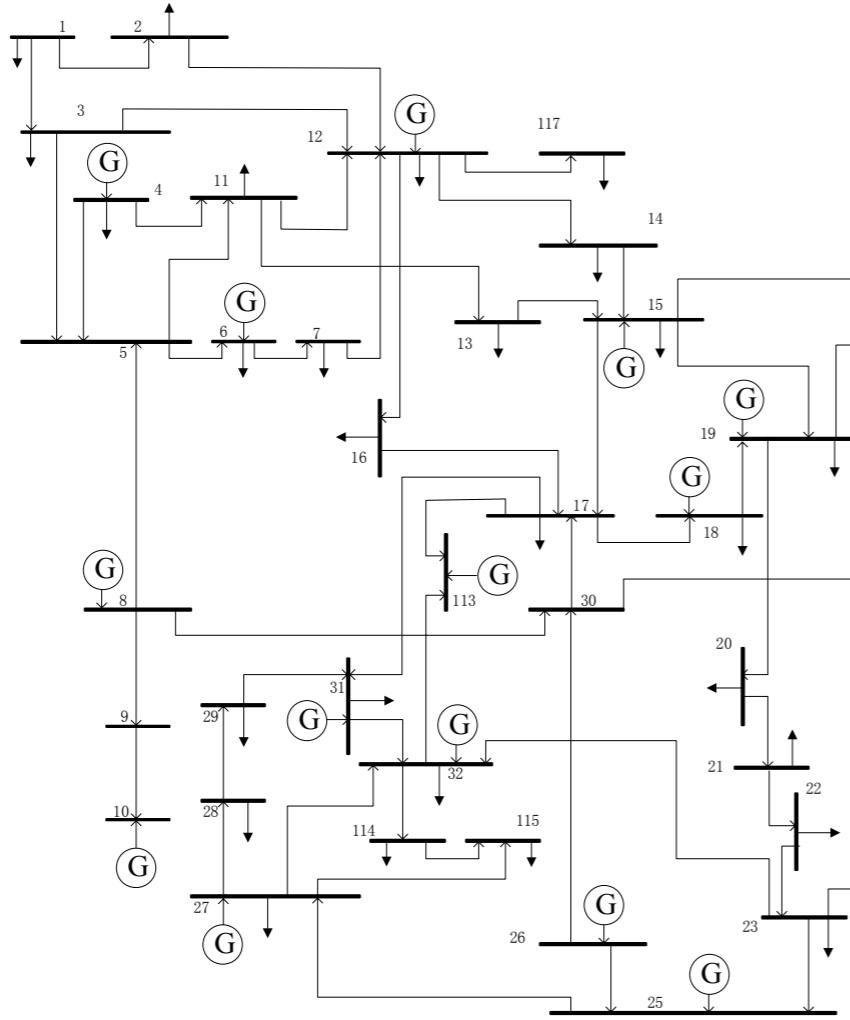


ACOPF Implementation (Pyomo & IPOPT)

Contingency-Constrained ACOPF (PySP)

Parallel Interior-point Methods (Schur-IPOPT)

AC Optimal Power Flow (ACOPF)



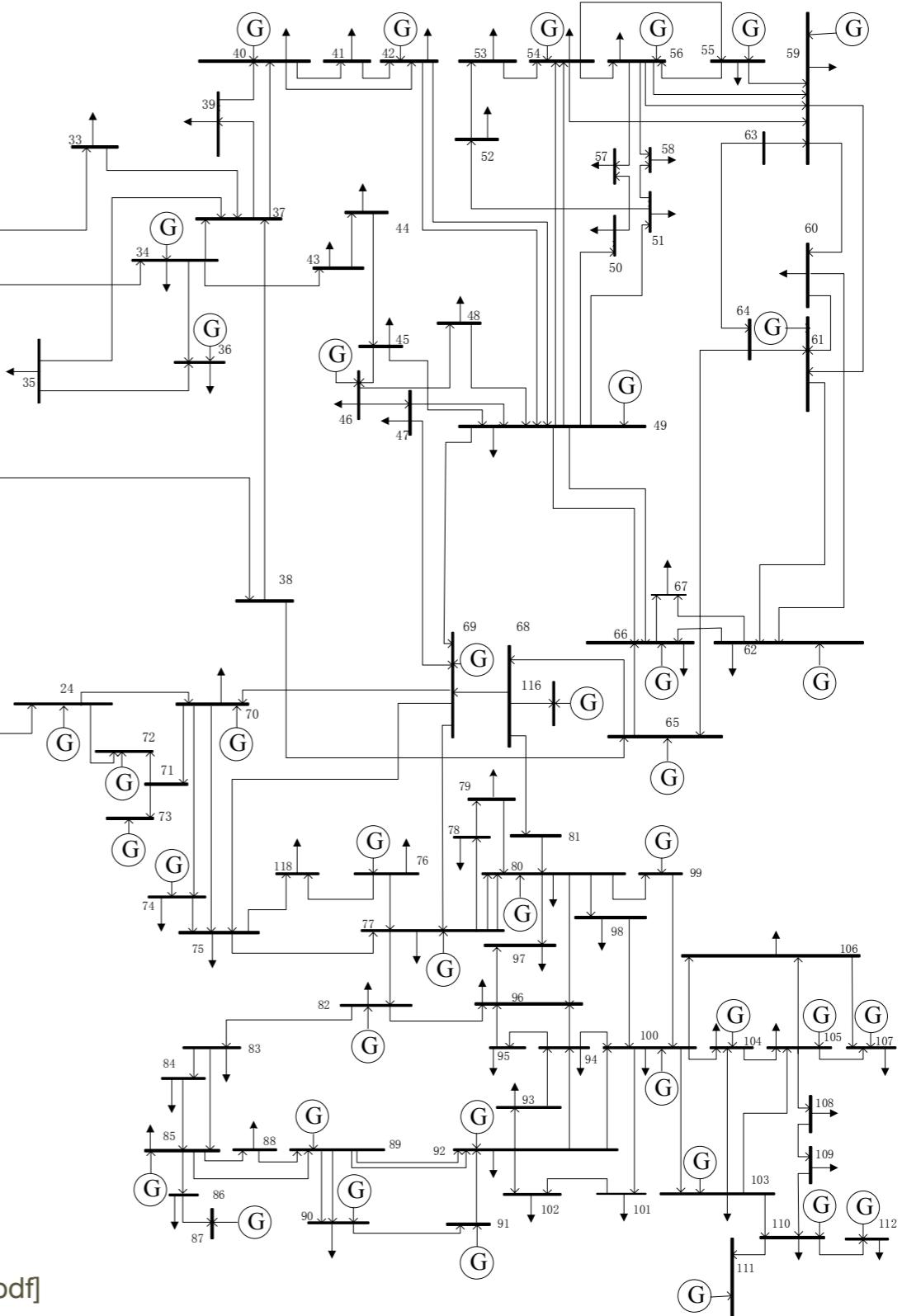
System Description:

118 buses
186 branches
91 load sides
54 thermal units

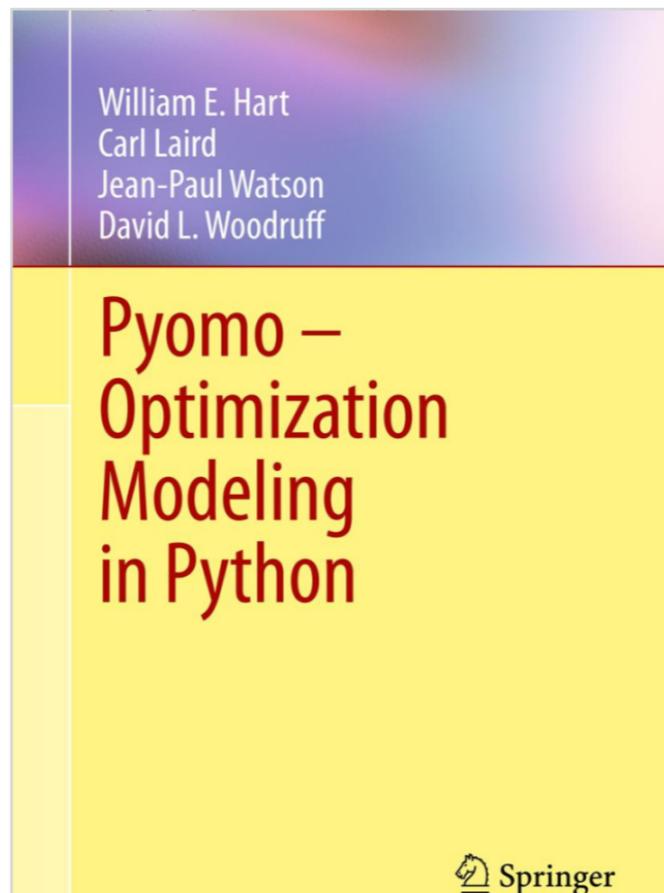
One-line Diagram of IEEE 118-bus Test System

IIT Power Group, 2003

[http://motor.ece.iit.edu/data/ltscc/IEEE118bus_figure.pdf]

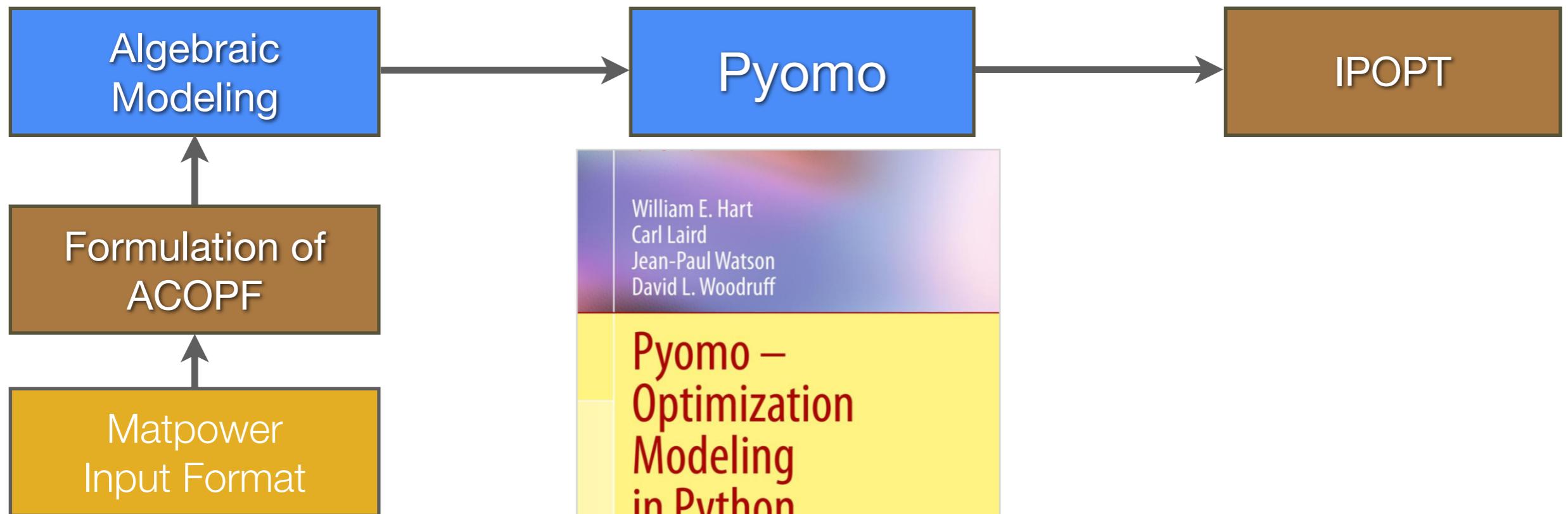


Building the model with Pyomo

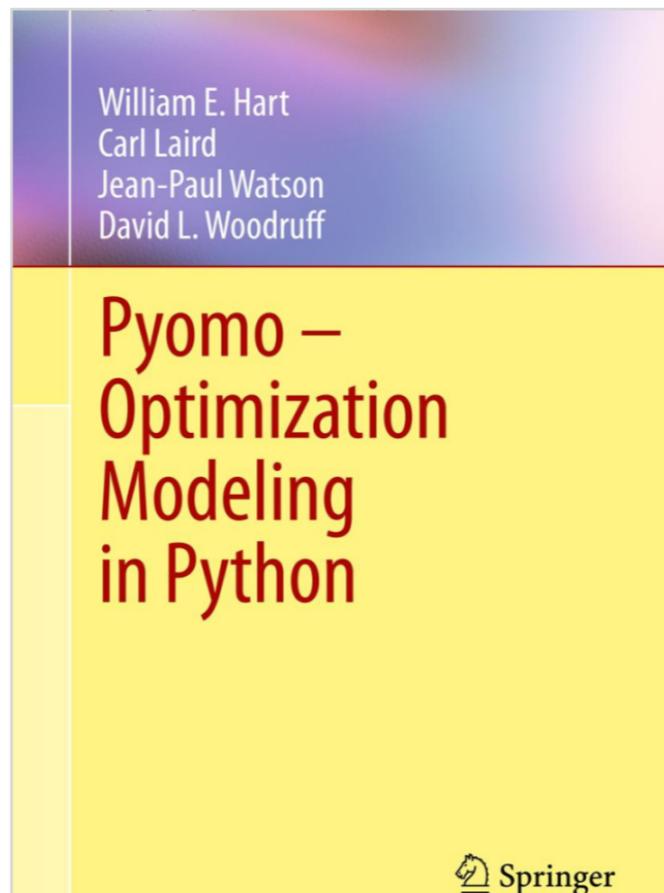


Hackebeil, Hart, Laird,
Siirola, Watson, Woodruff,
and many others...

Building the model with Pyomo



Integrated
Python-based
toolchain

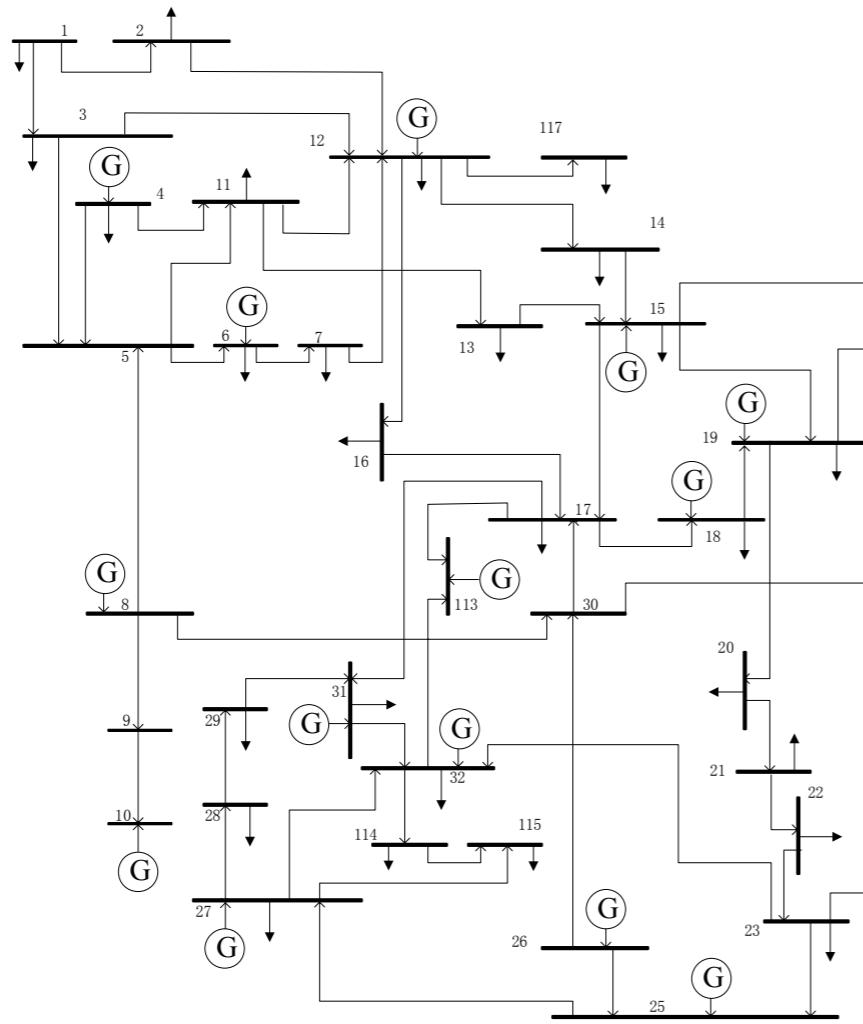


Hakebeil, Hart, Laird,
Siirola, Watson, Woodruff,
and many others...

ACOPF Solution with Pyomo and Ipopt

Case Name	Number of Variables	Δ Objective Function (\$) compared with Matpower
case30	46	2.410E-05
case9	95	-2.931E-04
case9Q	95	-4.806E-04
case6ww	105	-7.390E-05
case14	197	-9.430E-05
case30Q	399	-3.940E-05
case_ieee30	399	-8.082E-04
case24_ieee_rts	416	-5.421E-03
case39	465	2.282E-04
case57	767	8.225E-04
case118_laird	1,831	-1.020E-03
case2383wp	28,456	-2.939E-02
case2737sop	33,742	1.139E-03
case2736sp	33,807	-5.044E-03
case2746wop	34,013	-1.544E-04
case2746wp	34,063	-6.135E-03
case3012wp	35,242	-1.976E-02
case3120sp	36,247	-1.149E-02

N-1 Contingency-Constrained ACOPF Problem



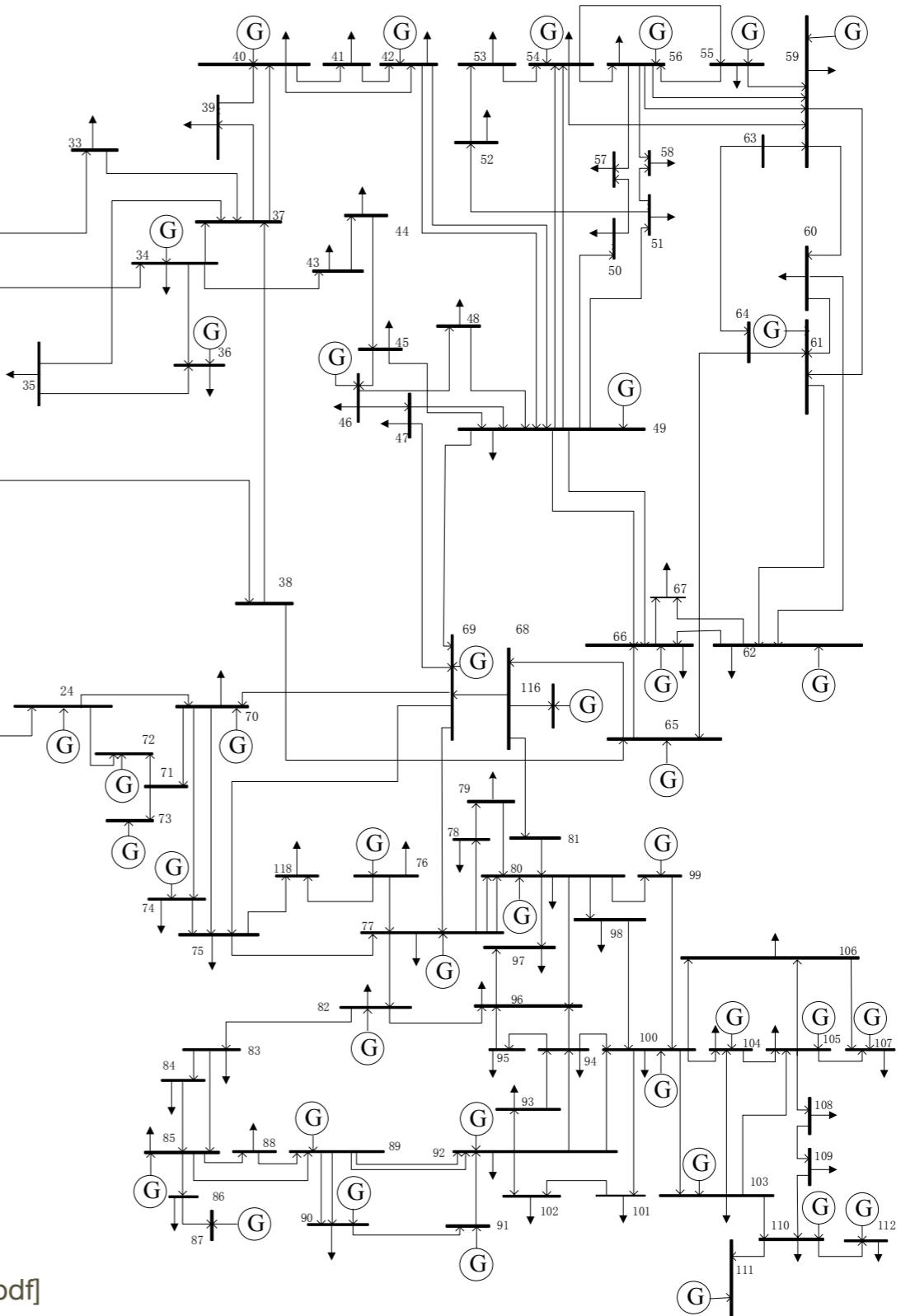
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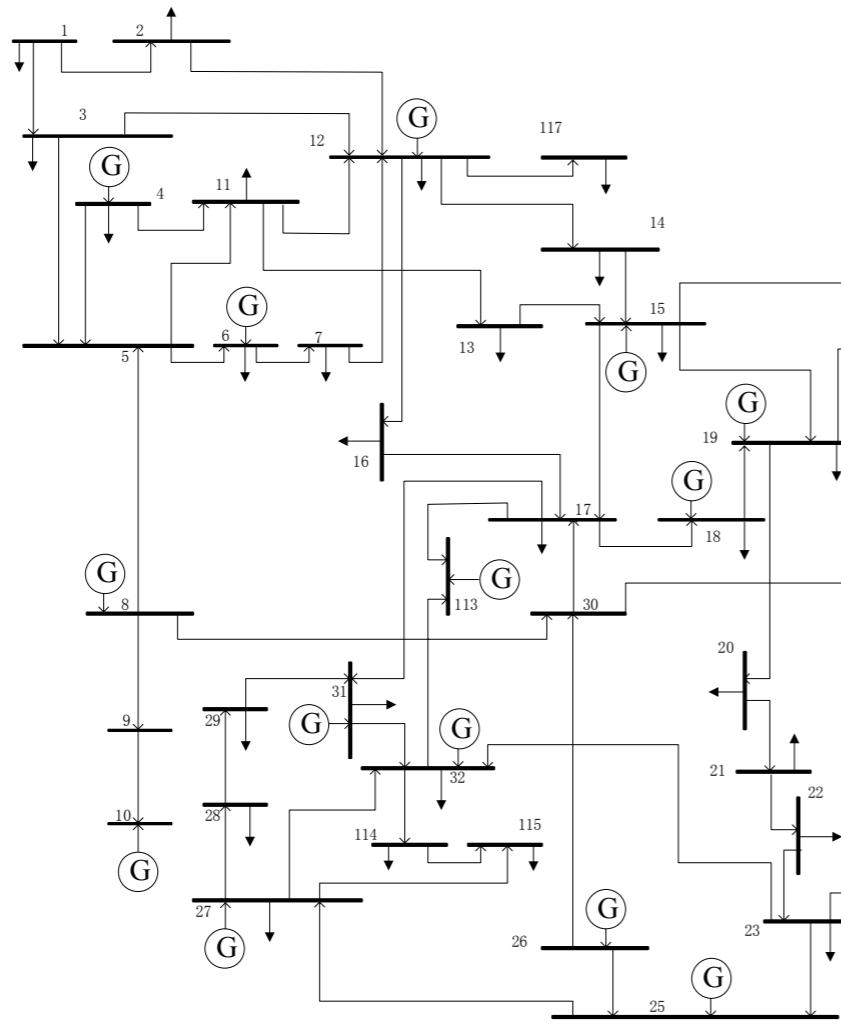
One-line Diagram of IEEE 118-bus Test System

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[http://motor.ece.iit.edu/data/ltscc/IEEE118bus_figure.pdf]



N-1 Contingency-Constrained ACOPF Problem



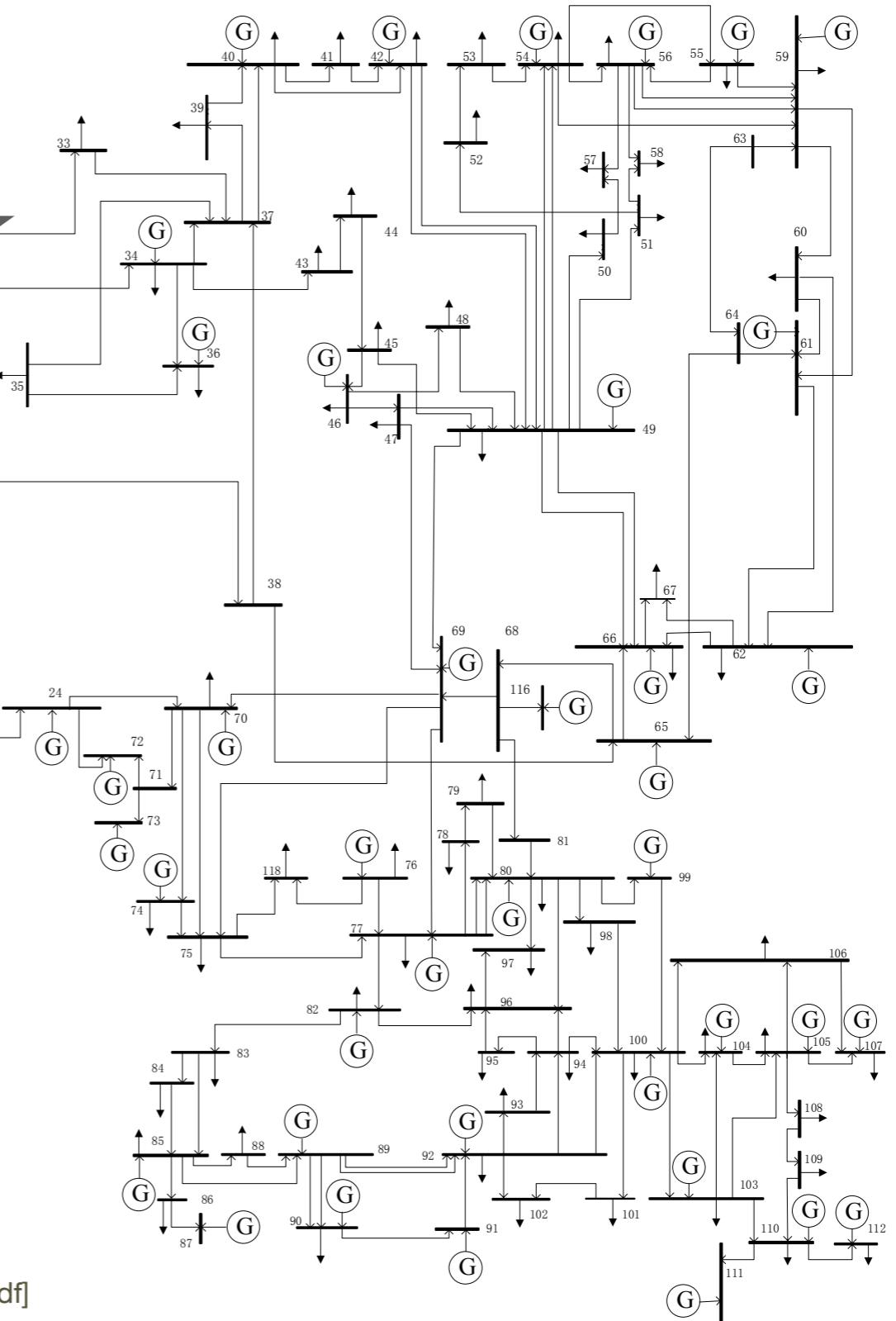
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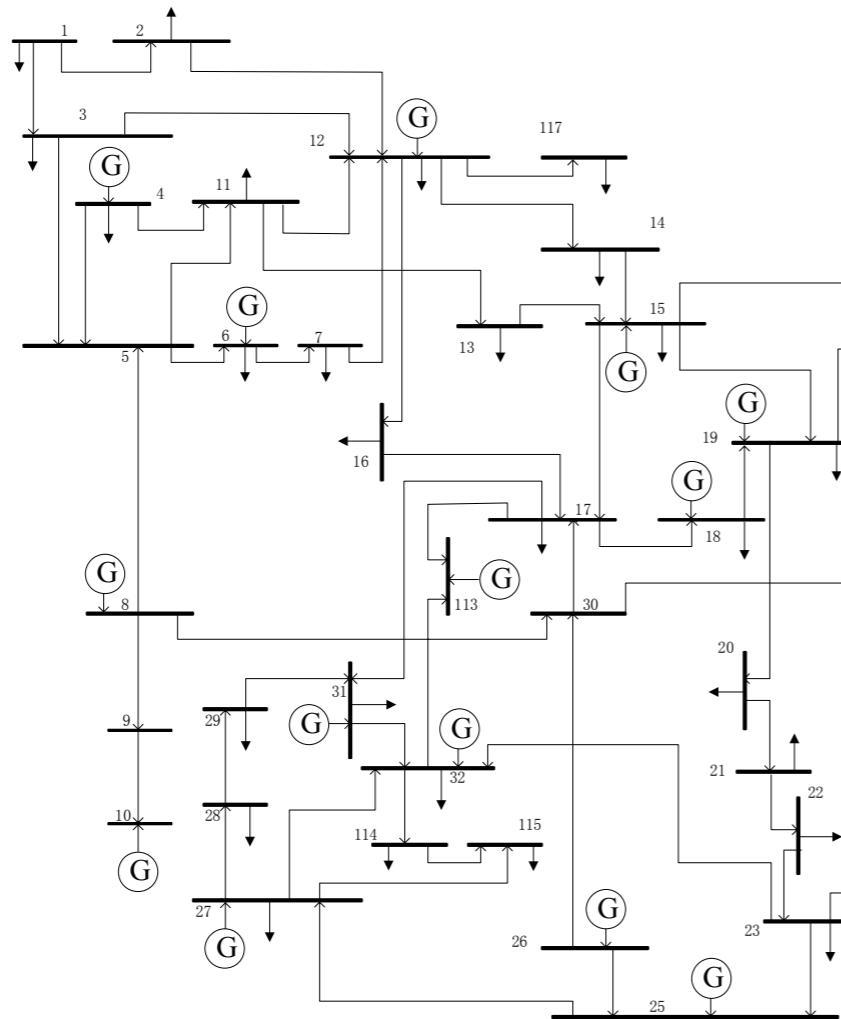
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N-1 Contingency-Constrained ACOPF Problem



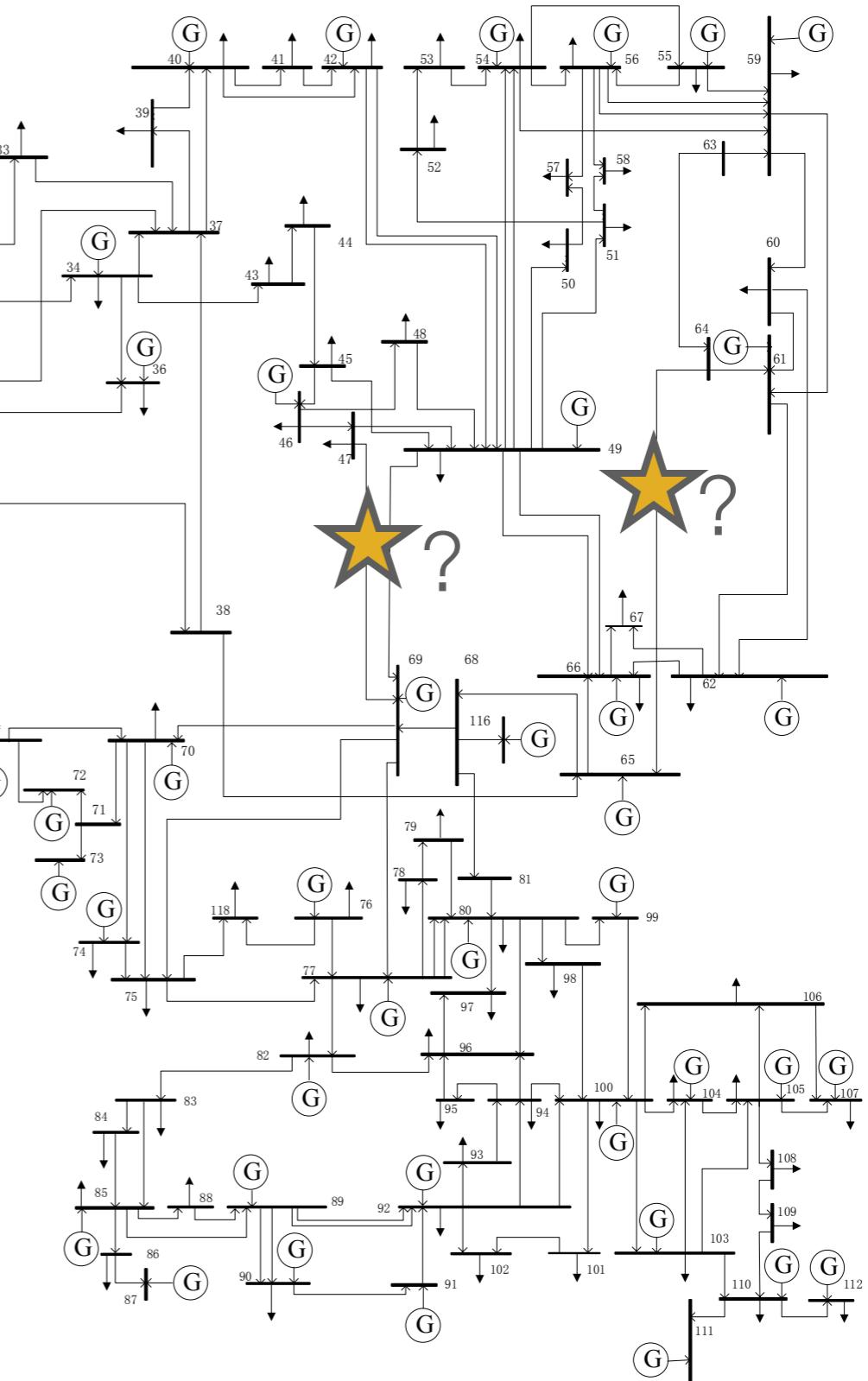
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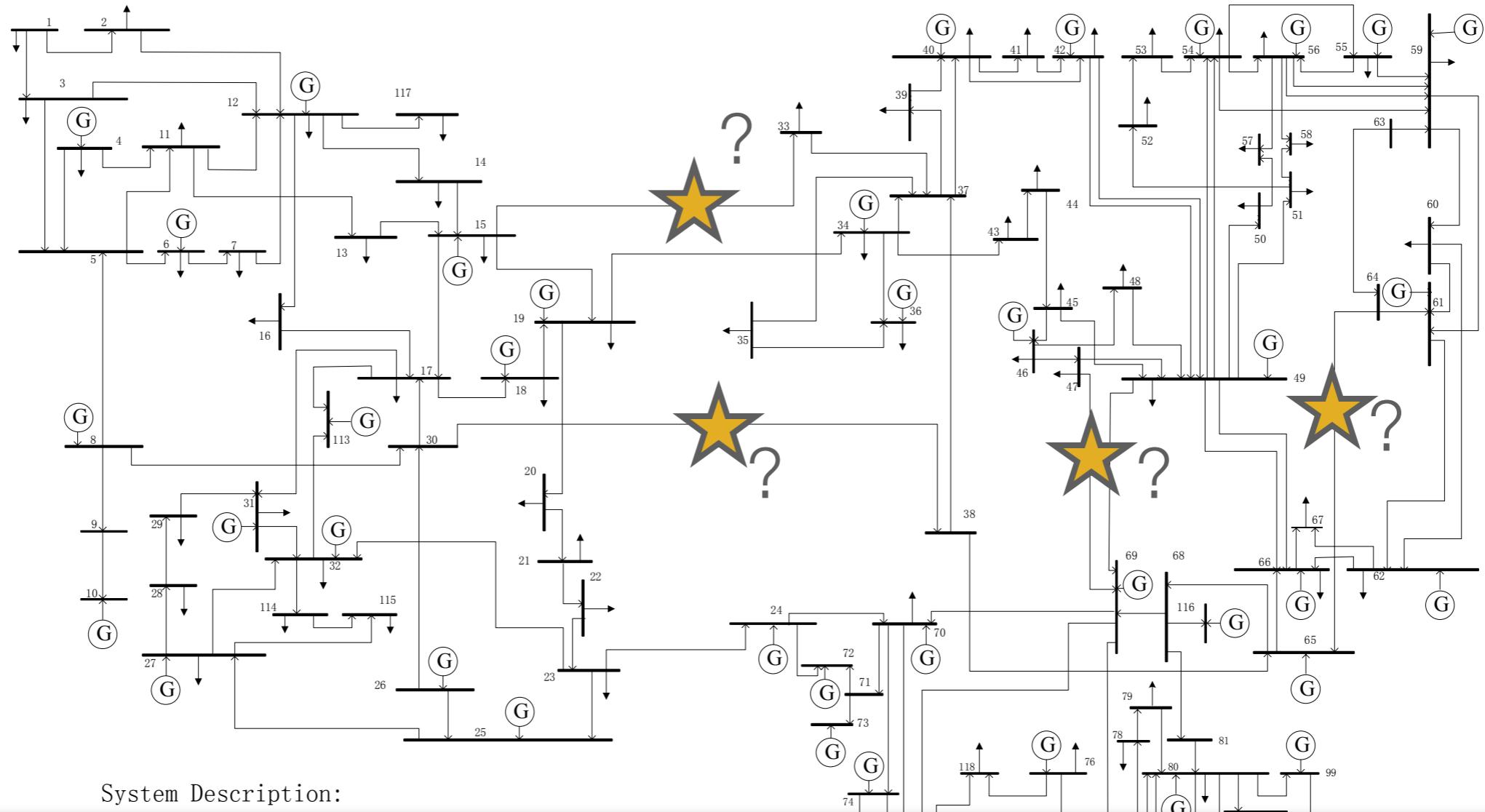
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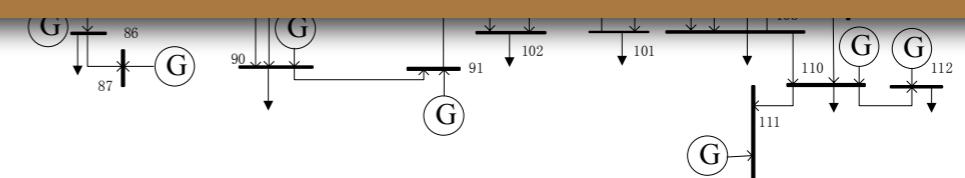
N-1 Contingency-Constrained ACOPF Problem



Make a decision about how to operate now while considering all N-1 possibilities for transmission element failure.

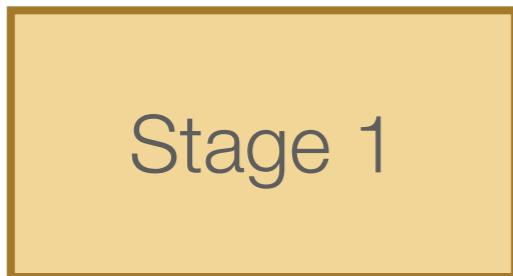
IIT Power Group, 2003

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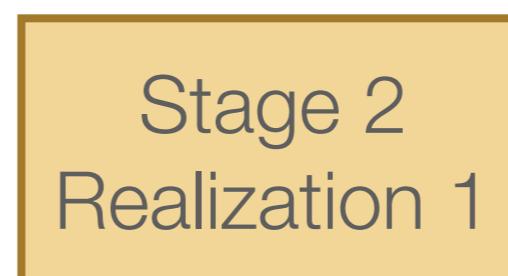
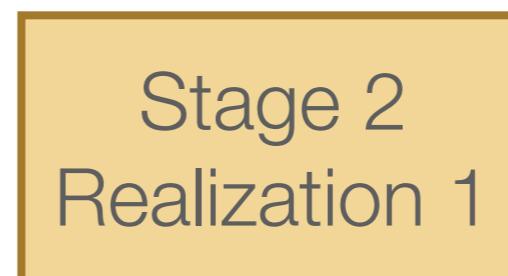


Two-Stage Stochastic Programming

Stage 1
(current operation)



Stage 2
(recourse)

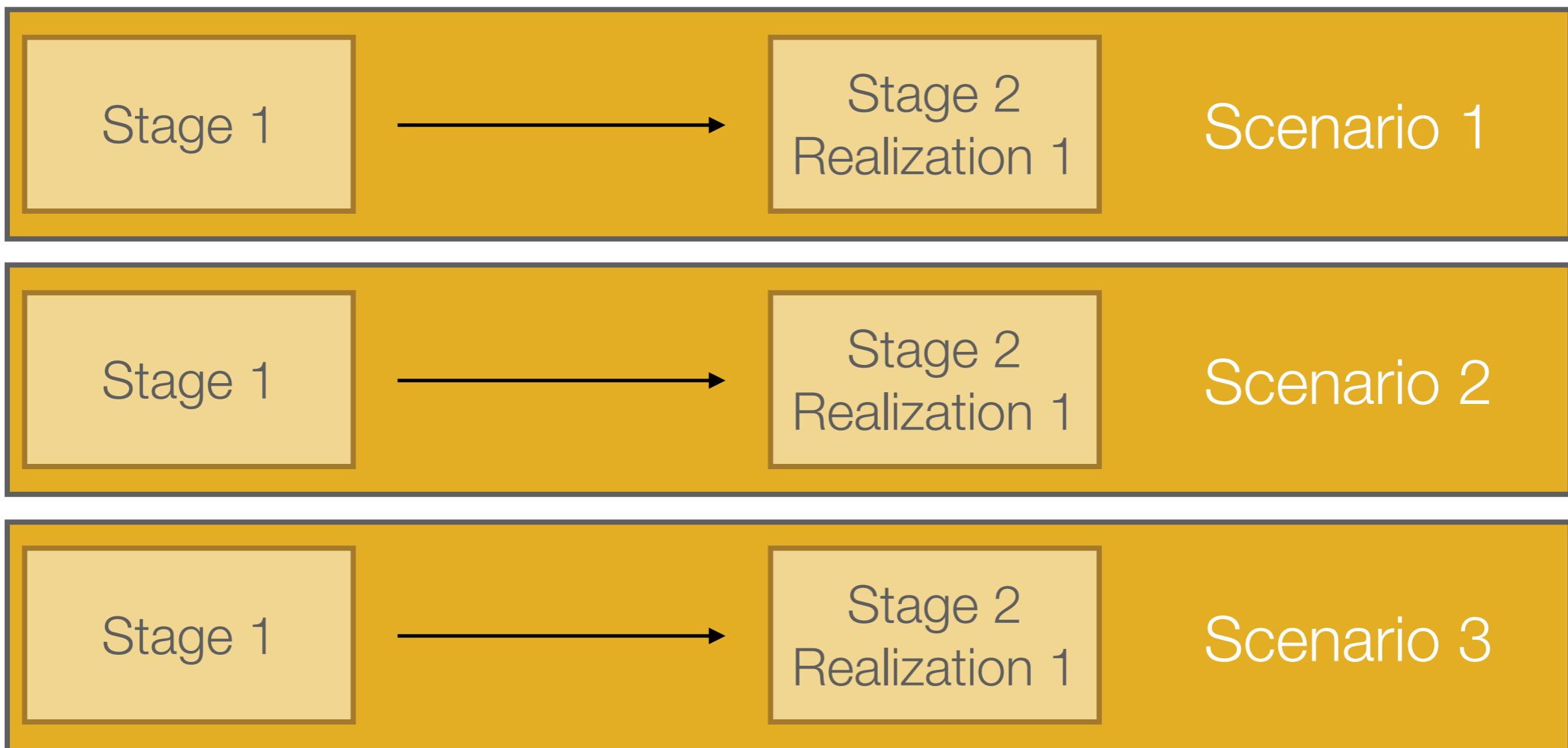


⋮

Two-Stage Stochastic Programming

Stage 1

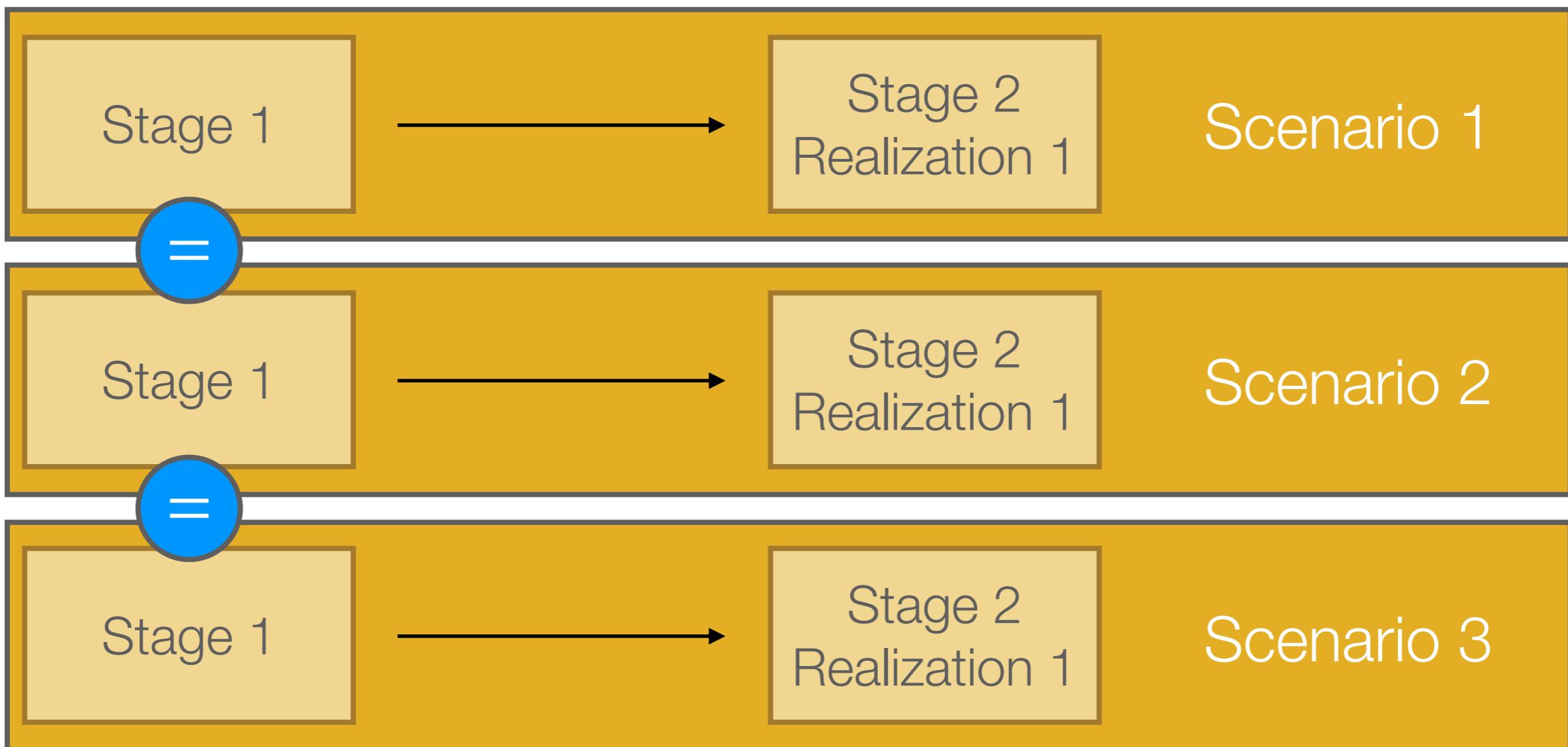
Stage 2



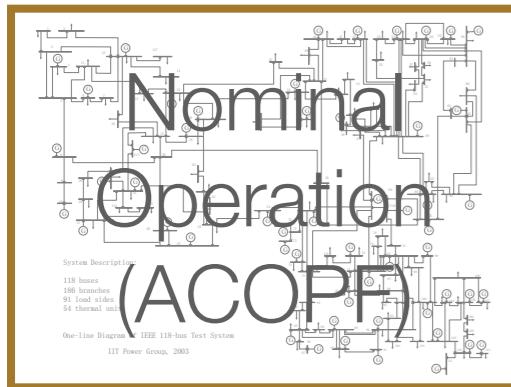
Two-Stage Stochastic Programming

Stage 1

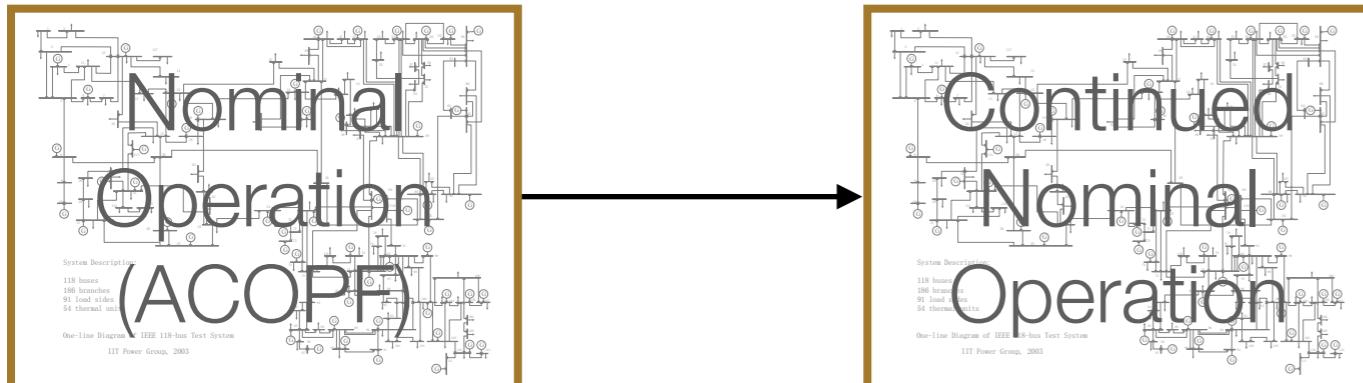
Stage 2



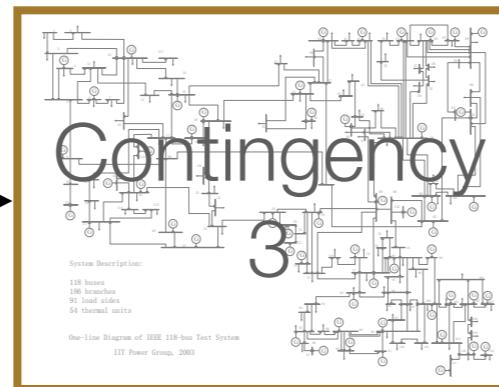
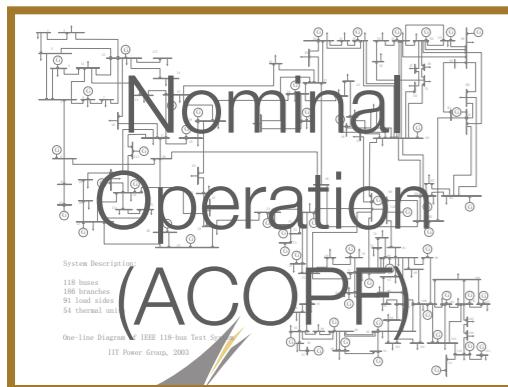
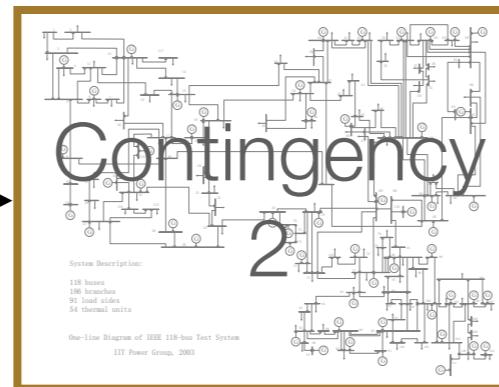
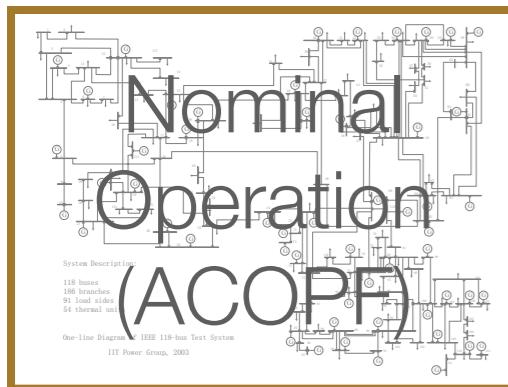
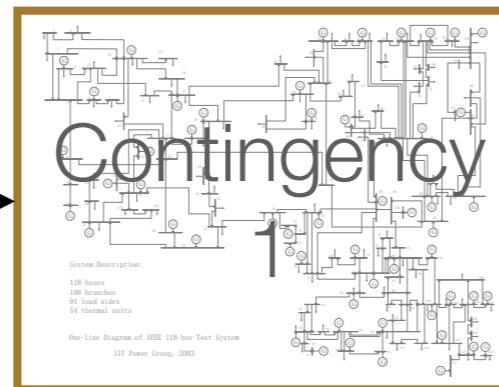
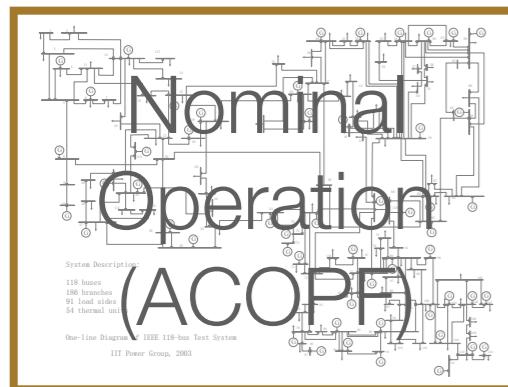
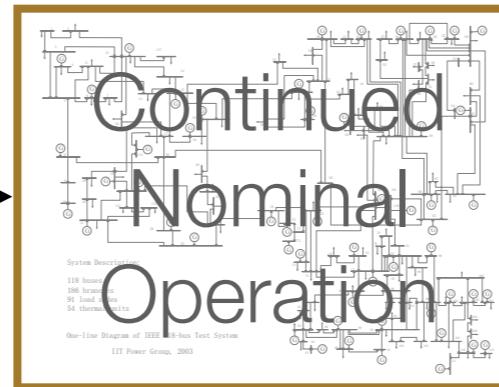
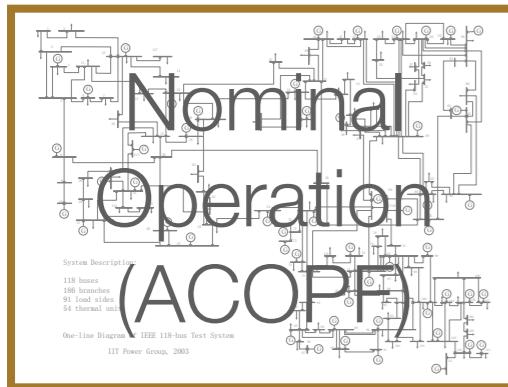
Contingency-Constrained ACOPF Problem



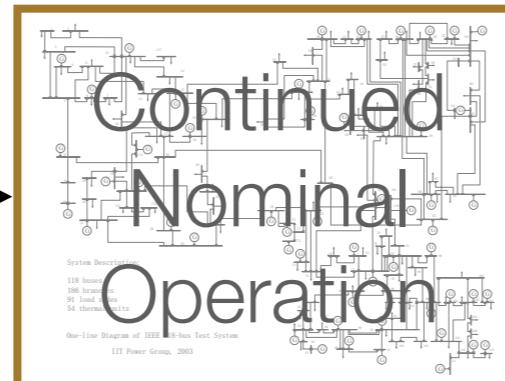
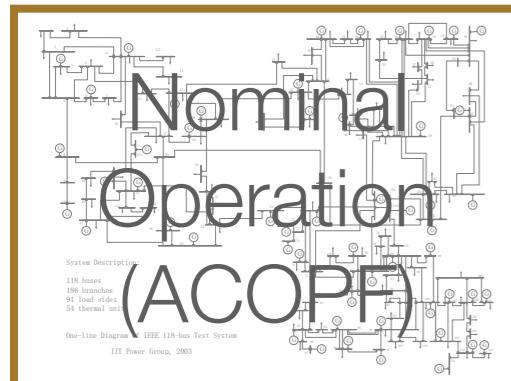
Contingency-Constrained ACOPF Problem



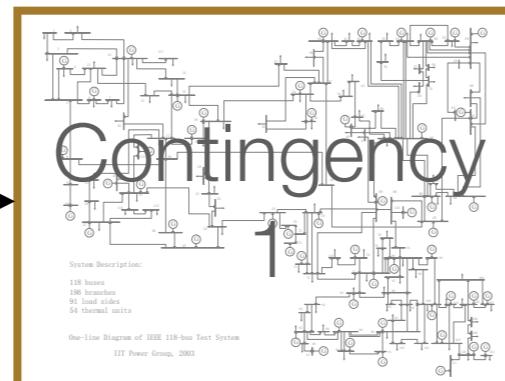
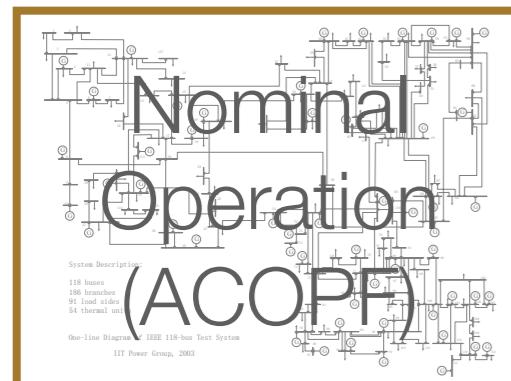
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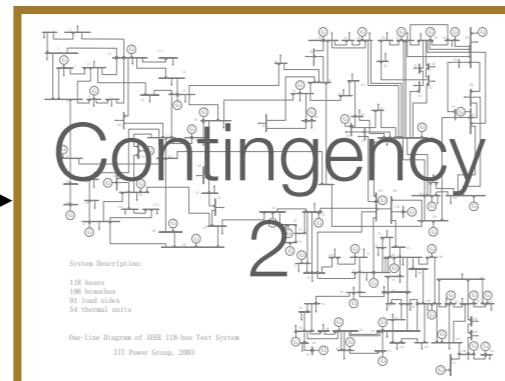
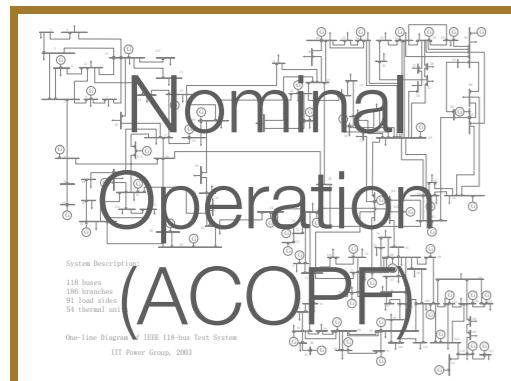
Contingency-Constrained ACOPF Problem



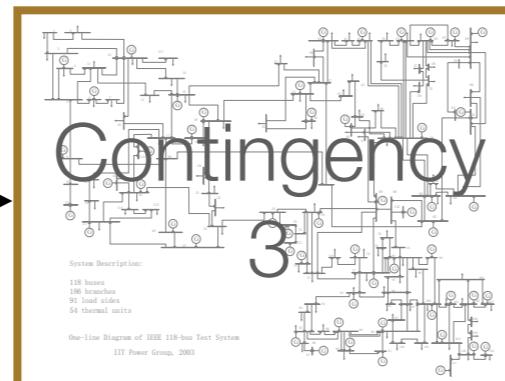
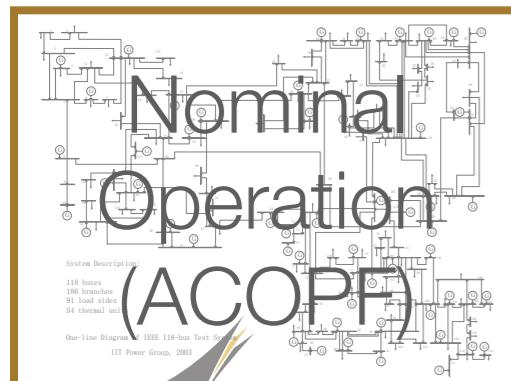
Nonlinear two-stage stochastic programming problem



AC power flow model for each scenario and stage



Penalty for inability to meet demands



Ramping constraints for changes in generator set points

Very large-scale nonlinear programming problem
(Millions of variables)

Contingency-Constrained Stochastic Programming ACOPF Formulation

\mathcal{L} Lines

\mathcal{B} All Buses

\mathcal{S} Scenarios

\mathcal{G} Generators

\mathcal{D} Load Buses

\mathcal{T} Time periods

$$\min \sum_{s \in S} p_s \sum_{t \in T} \left[\sum_{g \in \mathcal{G}} C_g^G(P_{g,t,s}^G, Q_{g,t,s}^G) + \rho_1 \sum_{g \in \mathcal{G}} \left[(P_{g,t,s}^G - P_{g,t,s}^{\star G})^2 + (Q_{g,t,s}^G - Q_{g,t,s}^{\star G})^2 \right] \right. \\ \left. + \rho_2 \sum_{b \in \mathcal{D}} \left[(P_{b,t,s}^L - P_b^{\star L})^2 + (Q_{b,t,s}^L - Q_b^{\star L})^2 \right] \right]$$

$$\text{s.t. } \begin{bmatrix} i_{l,t,s}^{fr} \\ i_{l,t,s}^{fj} \\ i_{l,t,s}^{tr} \\ i_{l,t,s}^{tj} \end{bmatrix} = Y_{l,t,s} \begin{bmatrix} v_{bf(l),t,s}^r \\ v_{bf(l),t,s}^j \\ v_{bt(l),t,s}^r \\ v_{bt(l),t,s}^j \end{bmatrix} \quad \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$$

$$P_{b,t,s}^S = Y_b^S \left[(v_{b,t,s}^r)^2 + (v_{b,t,s}^j)^2 \right] \quad \forall b \in \mathcal{H}, t \in \mathcal{T}, s \in \mathcal{S}$$

$$Q_{b,t,s}^S = -Y_b^S \left[(v_{b,t,s}^r)^2 + (v_{b,t,s}^j)^2 \right] \quad \forall b \in \mathcal{H}, t \in \mathcal{T}, s \in \mathcal{S}$$

$$0 = \sum_{l \in \mathcal{I}_b} P_{l,t,s}^t + \sum_{l \in \mathcal{O}_b} P_{l,t,s}^f + P_{b,t,s}^S + P_{b,t,s}^L - P_{b,t,s}^G \quad \forall b \in \mathcal{B}, t \in \mathcal{T}, s \in \mathcal{S}$$

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Objective: Expected Value of
Generator Op. Costs
+ Penalty

$$\text{s.t. } \begin{bmatrix} i_{l,t,s}^{fr} \\ i_{l,t,s}^{fj} \\ i_{l,t,s}^{tr} \\ i_{l,t,s}^{tj} \end{bmatrix} = Y_{l,t,s} \begin{bmatrix} v_{bf(l),t,s}^r \\ v_{bf(l),t,s}^j \\ v_{bt(l),t,s}^r \\ v_{bt(l),t,s}^j \end{bmatrix} \quad \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$$

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$$0 = \sum_{l \in \mathcal{I}_b} Q_{l,t,s}^t + \sum_{l \in \mathcal{O}_b} Q_{l,t,s}^f + Q_{b,t,s}^S + Q_{b,t,s}^L - Q_{b,t,s}^G \quad \forall b \in \mathcal{B}, t \in \mathcal{T}, s \in \mathcal{S}$$

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s.t.

$$\begin{bmatrix} i_{l,t,s}^{fr} \\ i_{l,t,s}^{fj} \\ i_{l,t,s}^{tr} \\ i_{l,t,s}^{tj} \end{bmatrix} = Y_{l,t,s} \begin{bmatrix} v_{bf(l),t,s}^r \\ v_{bf(l),t,s}^j \\ v_{bt(l),t,s}^r \\ v_{bt(l),t,s}^j \end{bmatrix} \quad \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$$

IV Relationships
(to/from) Every Line
(all S and T)

$$P_{b,t,s}^S = Y_b^S \left[(v_{b,t,s}^r)^2 + (v_{b,t,s}^j)^2 \right] \quad \forall b \in \mathcal{H}, t \in \mathcal{T}, s \in \mathcal{S}$$

PI Transmission

$$Q_{b,t,s}^S = -Y_b^S \left[(v_{b,t,s}^r)^2 + (v_{b,t,s}^j)^2 \right] \quad \forall b \in \mathcal{H}, t \in \mathcal{T}, s \in \mathcal{S}$$

Model

$$0 = \sum_{l \in \mathcal{I}_b} P_{l,t,s}^t + \sum_{l \in \mathcal{O}_b} P_{l,t,s}^f + P_{b,t,s}^S + P_{b,t,s}^L - P_{b,t,s}^G \quad \forall b \in \mathcal{B}, t \in \mathcal{T}, s \in \mathcal{S}$$

$$0 = \sum_{l \in \mathcal{I}_b} Q_{l,t,s}^t + \sum_{l \in \mathcal{O}_b} Q_{l,t,s}^f + Q_{b,t,s}^S + Q_{b,t,s}^L - Q_{b,t,s}^G \quad \forall b \in \mathcal{B}, t \in \mathcal{T}, s \in \mathcal{S}$$

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s.t.
$$\begin{bmatrix} i_{l,t,s}^{fr} \\ i_{l,t,s}^{fj} \\ i_{l,t,s}^{tr} \\ i_{l,t,s}^{tj} \end{bmatrix} = Y_{l,t,s} \begin{bmatrix} v_{bf(l),t,s}^r \\ v_{bf(l),t,s}^j \\ v_{bt(l),t,s}^r \\ v_{bt(l),t,s}^j \end{bmatrix} \quad \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$$

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Power Balance
(to/from) Every Bus
(all S and T)

Contingency-Constrained Stochastic Programming ACOPF Formulation

$P_{l,t,s}^f = v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fr} + v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	Power (to/from) Every Line (all S and T)
$Q_{l,t,s}^f = v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fr} - v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$P_{l,t,s}^t = v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tr} + v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$Q_{l,t,s}^t = v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tr} - v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$(P_{l,t,s}^f)^2 + (Q_{l,t,s}^f)^2 \leq (S^U)^2$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$(P_{l,t,s}^t)^2 + (Q_{l,t,s}^t)^2 \leq (S^U)^2$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$(V_b^L)^2 \leq (v_{b,t,s}^r)^2 + (v_{b,t,s}^j)^2 \leq (V_b^U)^2$	$\forall b \in \mathcal{B}, t \in \mathcal{T}, s \in \mathcal{S}$	
$P_g^{GL} \leq P_g^G \leq P_g^{GU}$	$\forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}$	
$Q_g^{GL} \leq Q_g^G \leq Q_g^{GU}$	$\forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}$	
$v_{rb,t,s}^j = 0$	$\forall t \in \mathcal{T}, s \in \mathcal{S}$	
$P_{b,1,s}^L = P_b^{\star L}$	$\forall b \in \mathcal{D}, s \in \mathcal{S}$	
$Q_{b,1,s}^L = Q_b^{\star L}$	$\forall b \in \mathcal{D}, s \in \mathcal{S}$	
$-P_g^{GR} \leq P_{g,1,s}^G - P_{g,2,s}^G \leq P_g^{GR}$	$\forall g \in \mathcal{G}, s \in \mathcal{S}$	
$P_{g,1,0}^G = P_{g,2,0}^G$	$\forall g \in \mathcal{G}$	
$Q_{g,1,0}^G = Q_{g,2,0}^G$	$\forall g \in \mathcal{G}$	
$P_{g,1,0}^G = P_{g,1,s}^G$	$\forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$	
$Q_{g,1,0}^G = Q_{g,1,s}^G$	$\forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$	

Contingency-Constrained Stochastic Programming ACOPF Formulation

$$P_{l,t,s}^f = v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fr} + v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fj} \quad \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$$

$$Q_{l,t,s}^f = v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fr} - v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fj} \quad \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$$

$$P_{l,t,s}^t = v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tr} + v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tj} \quad \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$$

$$Q_{l,t,s}^t = v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tr} - v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tj} \quad \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$$

$$(P_{l,t,s}^f)^2 + (Q_{l,t,s}^f)^2 \leq (S^U)^2 \quad \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$$

$$(P_{l,t,s}^t)^2 + (Q_{l,t,s}^t)^2 \leq (S^U)^2 \quad \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$$

$$(V_b^L)^2 \leq (v_{b,t,s}^r)^2 + (v_{b,t,s}^j)^2 \leq (V_b^U)^2 \quad \forall b \in \mathcal{B}, t \in \mathcal{T}, s \in \mathcal{S}$$

$$P_g^{GL} \leq P_g^G \leq P_g^{GU} \quad \forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}$$

$$Q_g^{GL} \leq Q_g^G \leq Q_g^{GU} \quad \forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}$$

$$v_{rb,t,s}^j = 0 \quad \forall t \in \mathcal{T}, s \in \mathcal{S}$$

$$P_{b,1,s}^L = P_b^{\star L} \quad \forall b \in \mathcal{D}, s \in \mathcal{S}$$

$$Q_{b,1,s}^L = Q_b^{\star L} \quad \forall b \in \mathcal{D}, s \in \mathcal{S}$$

$$-P_g^{GR} \leq P_{g,1,s}^G - P_{g,2,s}^G \leq P_g^{GR} \quad \forall g \in \mathcal{G}, s \in \mathcal{S}$$

$$P_{g,1,0}^G = P_{g,2,0}^G \quad \forall g \in \mathcal{G}$$

$$Q_{g,1,0}^G = Q_{g,2,0}^G \quad \forall g \in \mathcal{G}$$

$$P_{g,1,0}^G = P_{g,1,s}^G \quad \forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$$

$$Q_{g,1,0}^G = Q_{g,1,s}^G \quad \forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$$

Line limits on (to/from)
of every line
(all S and T)

Contingency-Constrained Stochastic Programming ACOPF Formulation

$$\begin{aligned}
 P_{l,t,s}^f &= v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fr} + v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fj} & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 Q_{l,t,s}^f &= v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fr} - v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fj} & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 P_{l,t,s}^t &= v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tr} + v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tj} & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 Q_{l,t,s}^t &= v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tr} - v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tj} & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 (P_{l,t,s}^f)^2 + (Q_{l,t,s}^f)^2 &\leq (S^U)^2 & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 (P_{l,t,s}^t)^2 + (Q_{l,t,s}^t)^2 &\leq (S^U)^2 & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 (V_b^L)^2 \leq (v_{b,t,s}^r)^2 + (v_{b,t,s}^j)^2 \leq (V_b^U)^2 & \forall b \in \mathcal{B}, t \in \mathcal{T}, s \in \mathcal{S}
 \end{aligned}$$

$P_g^{GL} \leq P_g^G \leq P_g^{GU}$	$\forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}$	Voltage limits on every bus (all S and T)
$Q_g^{GL} \leq Q_g^G \leq Q_g^{GU}$	$\forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}$	
$v_{rb,t,s}^j = 0$	$\forall t \in \mathcal{T}, s \in \mathcal{S}$	
$P_{b,1,s}^L = P_b^{\star L}$	$\forall b \in \mathcal{D}, s \in \mathcal{S}$	
$Q_{b,1,s}^L = Q_b^{\star L}$	$\forall b \in \mathcal{D}, s \in \mathcal{S}$	
$-P_g^{GR} \leq P_{g,1,s}^G - P_{g,2,s}^G \leq P_g^{GR}$	$\forall g \in \mathcal{G}, s \in \mathcal{S}$	
$P_{g,1,0}^G = P_{g,2,0}^G$	$\forall g \in \mathcal{G}$	
$Q_{g,1,0}^G = Q_{g,2,0}^G$	$\forall g \in \mathcal{G}$	
$P_{g,1,0}^G = P_{g,1,s}^G$	$\forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$	
$Q_{g,1,0}^G = Q_{g,1,s}^G$	$\forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$	

Contingency-Constrained Stochastic Programming ACOPF Formulation

$$\begin{aligned}
 P_{l,t,s}^f &= v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fr} + v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fj} & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 Q_{l,t,s}^f &= v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fr} - v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fj} & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 P_{l,t,s}^t &= v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tr} + v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tj} & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 Q_{l,t,s}^t &= v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tr} - v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tj} & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 (P_{l,t,s}^f)^2 + (Q_{l,t,s}^f)^2 &\leq (S^U)^2 & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 (P_{l,t,s}^t)^2 + (Q_{l,t,s}^t)^2 &\leq (S^U)^2 & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 (V_b^L)^2 \leq (v_{b,t,s}^r)^2 + (v_{b,t,s}^j)^2 &\leq (V_b^U)^2 & \forall b \in \mathcal{B}, t \in \mathcal{T}, s \in \mathcal{S}
 \end{aligned}$$

$$\begin{aligned}
 P_g^{GL} \leq P_g^G \leq P_g^{GU} && \forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S} \\
 Q_g^{GL} \leq Q_g^G \leq Q_g^{GU} && \forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}
 \end{aligned}$$

$$\begin{aligned}
 v_{rb,t,s}^j &= 0 & \forall t \in \mathcal{T}, s \in \mathcal{S} & \text{Generator bounds} \\
 P_{b,1,s}^L &= P_b^{\star L} & \forall b \in \mathcal{D}, s \in \mathcal{S} & \text{on every generator} \\
 Q_{b,1,s}^L &= Q_b^{\star L} & \forall b \in \mathcal{D}, s \in \mathcal{S} & \text{(all S and T)} \\
 -P_g^{GR} \leq P_{g,1,s}^G - P_{g,2,s}^G &\leq P_g^{GR} & \forall g \in \mathcal{G}, s \in \mathcal{S} \\
 P_{g,1,0}^G &= P_{g,2,0}^G & \forall g \in \mathcal{G} \\
 Q_{g,1,0}^G &= Q_{g,2,0}^G & \forall g \in \mathcal{G} \\
 P_{g,1,0}^G &= P_{g,1,s}^G & \forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\} \\
 Q_{g,1,0}^G &= Q_{g,1,s}^G & \forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}
 \end{aligned}$$

Contingency-Constrained Stochastic Programming ACOPF Formulation

$$\begin{aligned}
 P_{l,t,s}^f &= v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fr} + v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fj} & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 Q_{l,t,s}^f &= v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fr} - v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fj} & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 P_{l,t,s}^t &= v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tr} + v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tj} & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 Q_{l,t,s}^t &= v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tr} - v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tj} & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 (P_{l,t,s}^f)^2 + (Q_{l,t,s}^f)^2 &\leq (S^U)^2 & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 (P_{l,t,s}^t)^2 + (Q_{l,t,s}^t)^2 &\leq (S^U)^2 & \forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S} \\
 (V_b^L)^2 \leq (v_{b,t,s}^r)^2 + (v_{b,t,s}^j)^2 &\leq (V_b^U)^2 & \forall b \in \mathcal{B}, t \in \mathcal{T}, s \in \mathcal{S} \\
 P_g^{GL} \leq P_g^G \leq P_g^{GU} && \forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S} \\
 Q_g^{GL} \leq Q_g^G \leq Q_g^{GU} && \forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S} \\
 v_{rb,t,s}^j = 0 && \forall t \in \mathcal{T}, s \in \mathcal{S}
 \end{aligned}$$

$P_{b,1,s}^L = P_b^{\star L}$	$\forall b \in \mathcal{D}, s \in \mathcal{S}$	Reference Bus Constraint (all S and T)
$Q_{b,1,s}^L = Q_b^{\star L}$	$\forall b \in \mathcal{D}, s \in \mathcal{S}$	
$-P_g^{GR} \leq P_{g,1,s}^G - P_{g,2,s}^G \leq P_g^{GR}$	$\forall g \in \mathcal{G}, s \in \mathcal{S}$	
$P_{g,1,0}^G = P_{g,2,0}^G$	$\forall g \in \mathcal{G}$	
$Q_{g,1,0}^G = Q_{g,2,0}^G$	$\forall g \in \mathcal{G}$	
$P_{g,1,0}^G = P_{g,1,s}^G$	$\forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$	
$Q_{g,1,0}^G = Q_{g,1,s}^G$	$\forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$	

Contingency-Constrained Stochastic Programming ACOPF Formulation

$P_{l,t,s}^f = v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fr} + v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$Q_{l,t,s}^f = v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fr} - v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$P_{l,t,s}^t = v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tr} + v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$Q_{l,t,s}^t = v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tr} - v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$(P_{l,t,s}^f)^2 + (Q_{l,t,s}^f)^2 \leq (S^U)^2$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$(P_{l,t,s}^t)^2 + (Q_{l,t,s}^t)^2 \leq (S^U)^2$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$(V_b^L)^2 \leq (v_{b,t,s}^r)^2 + (v_{b,t,s}^j)^2 \leq (V_b^U)^2$	$\forall b \in \mathcal{B}, t \in \mathcal{T}, s \in \mathcal{S}$	
$P_g^{GL} \leq P_g^G \leq P_g^{GU}$	$\forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}$	Fix load values for Nominal stage (all S)
$Q_g^{GL} \leq Q_g^G \leq Q_g^{GU}$	$\forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}$	
$v_{rb,t,s}^j = 0$	$\forall t \in \mathcal{T}, s \in \mathcal{S}$	
$P_{b,1,s}^L = P_b^{\star L}$	$\forall b \in \mathcal{D}, s \in \mathcal{S}$	
$Q_{b,1,s}^L = Q_b^{\star L}$	$\forall b \in \mathcal{D}, s \in \mathcal{S}$	
$-P_g^{GR} \leq P_{g,1,s}^G - P_{g,2,s}^G \leq P_g^{GR}$	$\forall g \in \mathcal{G}, s \in \mathcal{S}$	
$P_{g,1,0}^G = P_{g,2,0}^G$	$\forall g \in \mathcal{G}$	
$Q_{g,1,0}^G = Q_{g,2,0}^G$	$\forall g \in \mathcal{G}$	
$P_{g,1,0}^G = P_{g,1,s}^G$	$\forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$	
$Q_{g,1,0}^G = Q_{g,1,s}^G$	$\forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$	

Contingency-Constrained Stochastic Programming ACOPF Formulation

$P_{l,t,s}^f = v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fr} + v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$
$Q_{l,t,s}^f = v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fr} - v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$
$P_{l,t,s}^t = v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tr} + v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$
$Q_{l,t,s}^t = v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tr} - v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$
$(P_{l,t,s}^f)^2 + (Q_{l,t,s}^f)^2 \leq (S^U)^2$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$
$(P_{l,t,s}^t)^2 + (Q_{l,t,s}^t)^2 \leq (S^U)^2$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$
$(V_b^L)^2 \leq (v_{b,t,s}^r)^2 + (v_{b,t,s}^j)^2 \leq (V_b^U)^2$	$\forall b \in \mathcal{B}, t \in \mathcal{T}, s \in \mathcal{S}$
$P_g^{GL} \leq P_g^G \leq P_g^{GU}$	$\forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}$
$Q_g^{GL} \leq Q_g^G \leq Q_g^{GU}$	$\forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}$
$v_{rb,t,s}^j = 0$	$\forall t \in \mathcal{T}, s \in \mathcal{S}$
$P_{b,1,s}^L = P_b^{\star L}$	$\forall b \in \mathcal{D}, s \in \mathcal{S}$
$Q_{b,1,s}^L = Q_b^{\star L}$	$\forall b \in \mathcal{D}, s \in \mathcal{S}$
$-P_g^{GR} \leq P_{g,1,s}^G - P_{g,2,s}^G \leq P_g^{GR}$	$\forall g \in \mathcal{G}, s \in \mathcal{S}$
$P_{g,1,0}^G = P_{g,2,0}^G$	$\forall g \in \mathcal{G}$
$Q_{g,1,0}^G = Q_{g,2,0}^G$	$\forall g \in \mathcal{G}$
$P_{g,1,0}^G = P_{g,1,s}^G$	$\forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$
$Q_{g,1,0}^G = Q_{g,1,s}^G$	$\forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$

Ramping constraint
on all generators
from stage 1 to stage 2
(all S)

Contingency-Constrained Stochastic Programming ACOPF Formulation

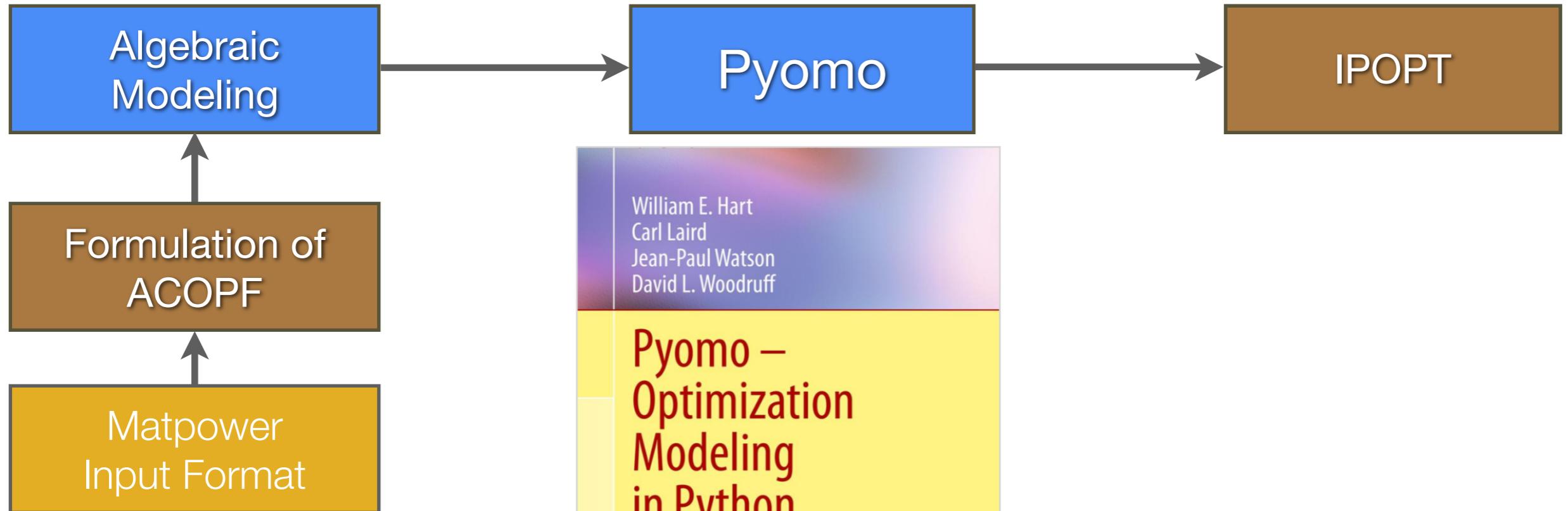
$P_{l,t,s}^f = v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fr} + v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$Q_{l,t,s}^f = v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fr} - v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$P_{l,t,s}^t = v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tr} + v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$Q_{l,t,s}^t = v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tr} - v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$(P_{l,t,s}^f)^2 + (Q_{l,t,s}^f)^2 \leq (S^U)^2$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$(P_{l,t,s}^t)^2 + (Q_{l,t,s}^t)^2 \leq (S^U)^2$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$	
$(V_b^L)^2 \leq (v_{b,t,s}^r)^2 + (v_{b,t,s}^j)^2 \leq (V_b^U)^2$	$\forall b \in \mathcal{B}, t \in \mathcal{T}, s \in \mathcal{S}$	
$P_g^{GL} \leq P_g^G \leq P_g^{GU}$	$\forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}$	
$Q_g^{GL} \leq Q_g^G \leq Q_g^{GU}$	$\forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}$	
$v_{rb,t,s}^j = 0$	$\forall t \in \mathcal{T}, s \in \mathcal{S}$	Enforce generator power solution equal in stage 1 and 2 for Nominal to Nominal
$P_{b,1,s}^L = P_b^{\star L}$	$\forall b \in \mathcal{D}, s \in \mathcal{S}$	
$Q_{b,1,s}^L = Q_b^{\star L}$	$\forall b \in \mathcal{D}, s \in \mathcal{S}$	
$-P_g^{GR} \leq P_{g,1,s}^G - P_{g,2,s}^G \leq P_g^{GR}$	$\forall g \in \mathcal{G}, s \in \mathcal{S}$	
$P_{g,1,0}^G = P_{g,2,0}^G$	$\forall g \in \mathcal{G}$	
$Q_{g,1,0}^G = Q_{g,2,0}^G$	$\forall g \in \mathcal{G}$	
$P_{g,1,0}^G = P_{g,1,s}^G$	$\forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$	
$Q_{g,1,0}^G = Q_{g,1,s}^G$	$\forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$	

Contingency-Constrained Stochastic Programming ACOPF Formulation

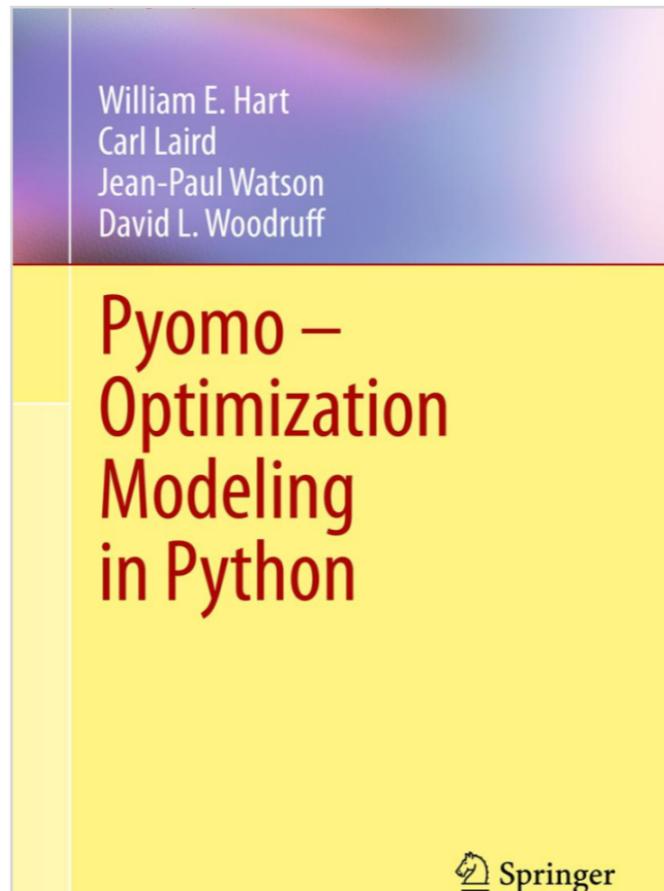
$P_{l,t,s}^f = v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fr} + v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$
$Q_{l,t,s}^f = v_{bf(l),t,s}^j \cdot i_{l,t,s}^{fr} - v_{bf(l),t,s}^r \cdot i_{l,t,s}^{fj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$
$P_{l,t,s}^t = v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tr} + v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$
$Q_{l,t,s}^t = v_{bt(l),t,s}^j \cdot i_{l,t,s}^{tr} - v_{bt(l),t,s}^r \cdot i_{l,t,s}^{tj}$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$
$(P_{l,t,s}^f)^2 + (Q_{l,t,s}^f)^2 \leq (S^U)^2$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$
$(P_{l,t,s}^t)^2 + (Q_{l,t,s}^t)^2 \leq (S^U)^2$	$\forall l \in \mathcal{L}, t \in \mathcal{T}, s \in \mathcal{S}$
$(V_b^L)^2 \leq (v_{b,t,s}^r)^2 + (v_{b,t,s}^j)^2 \leq (V_b^U)^2$	$\forall b \in \mathcal{B}, t \in \mathcal{T}, s \in \mathcal{S}$
$P_g^{GL} \leq P_g^G \leq P_g^{GU}$	$\forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}$
$Q_g^{GL} \leq Q_g^G \leq Q_g^{GU}$	$\forall g \in \mathcal{G}, t \in \mathcal{T}, s \in \mathcal{S}$
$v_{rb,t,s}^j = 0$	$\forall t \in \mathcal{T}, s \in \mathcal{S}$
$P_{b,1,s}^L = P_b^{\star L}$	$\forall b \in \mathcal{D}, s \in \mathcal{S}$
$Q_{b,1,s}^L = Q_b^{\star L}$	$\forall b \in \mathcal{D}, s \in \mathcal{S}$
$-P_g^{GR} \leq P_{g,1,s}^G - P_{g,2,s}^G \leq P_g^{GR}$	$\forall g \in \mathcal{G}, s \in \mathcal{S}$
$P_{g,1,0}^G = P_{g,2,0}^G$	$\forall g \in \mathcal{G}$
$Q_{g,1,0}^G = Q_{g,2,0}^G$	$\forall g \in \mathcal{G}$
$P_{g,1,0}^G = P_{g,1,s}^G$	$\forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$
$Q_{g,1,0}^G = Q_{g,1,s}^G$	$\forall g \in \mathcal{G}, s \in \mathcal{S}/\{0\}$

Non-anticipativity
constraints

Building the model with Pyomo and PySP

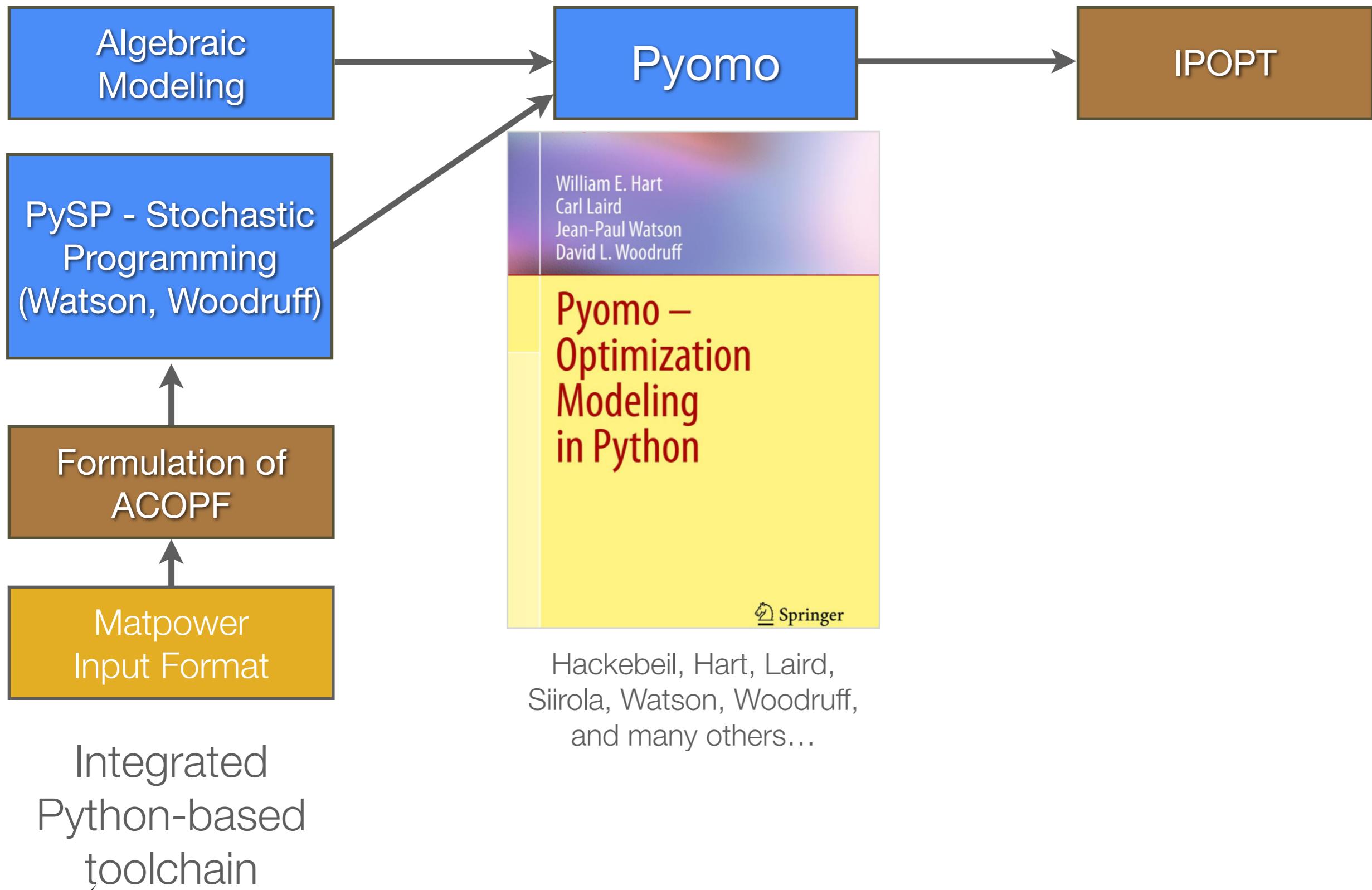


Integrated
Python-based
toolchain

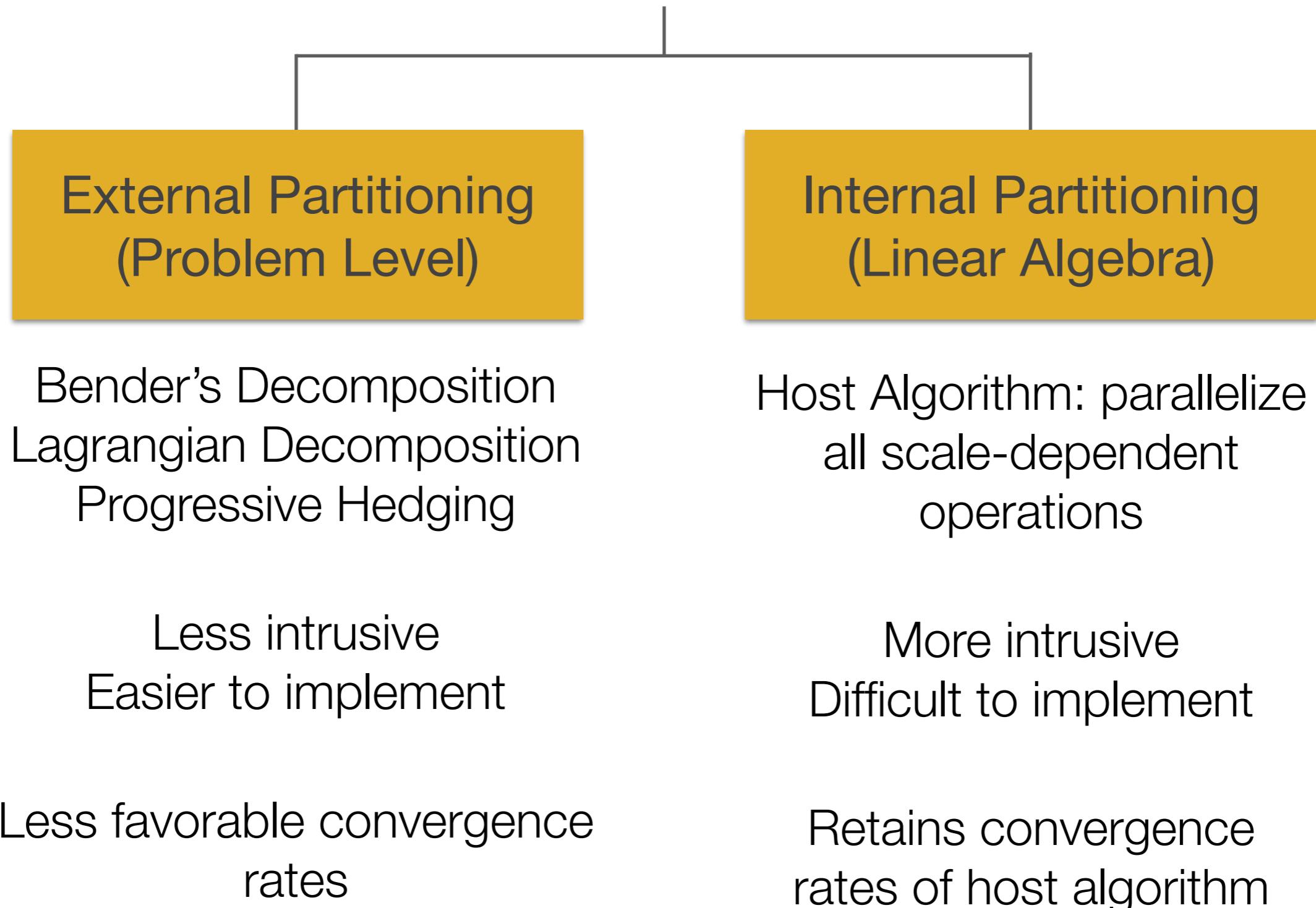


Hackebeil, Hart, Laird,
Siirola, Watson, Woodruff,
and many others...

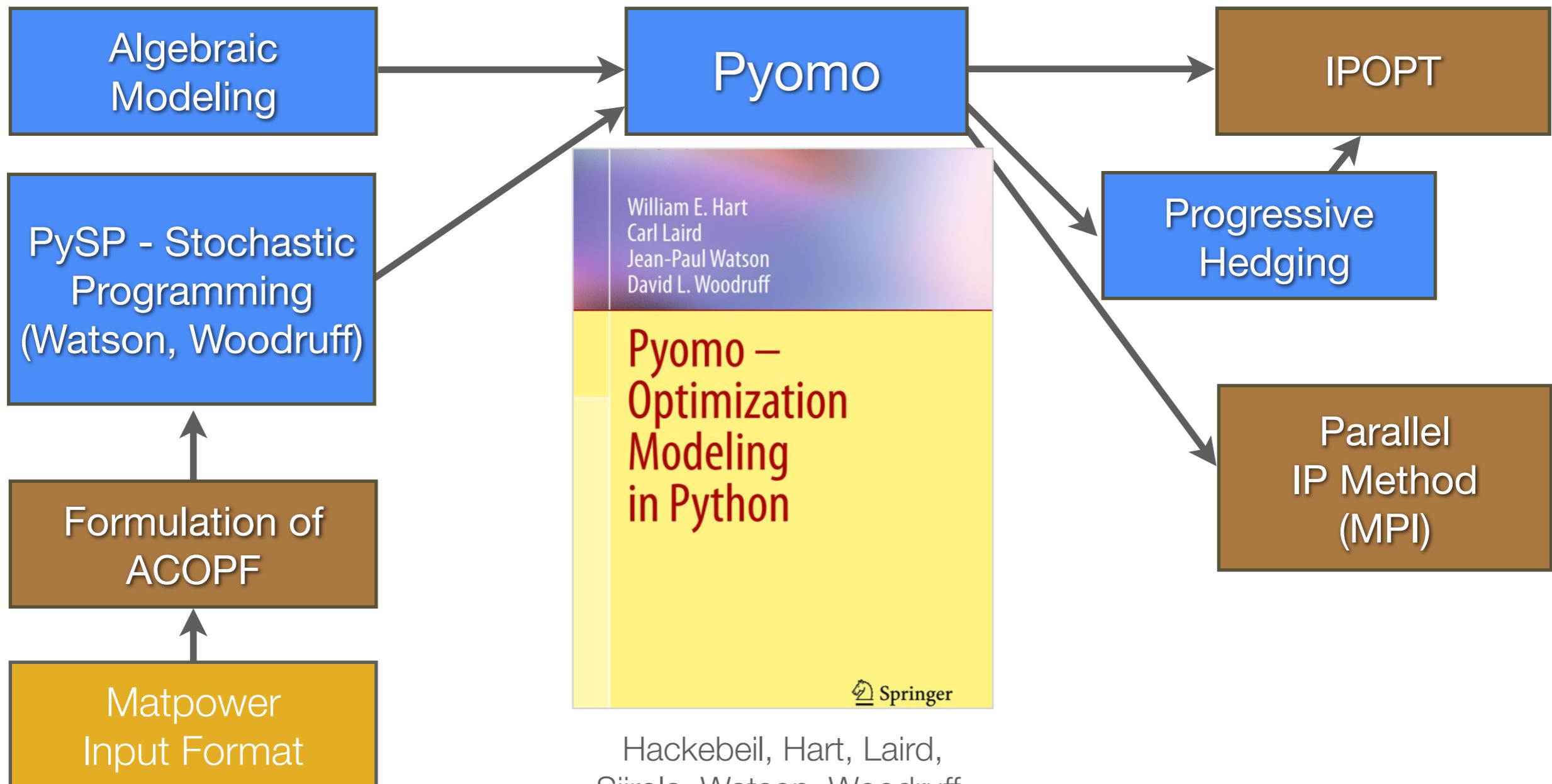
Building the model with Pyomo and PySP



Parallel Algorithms for Large-Scale Nonlinear Optimization

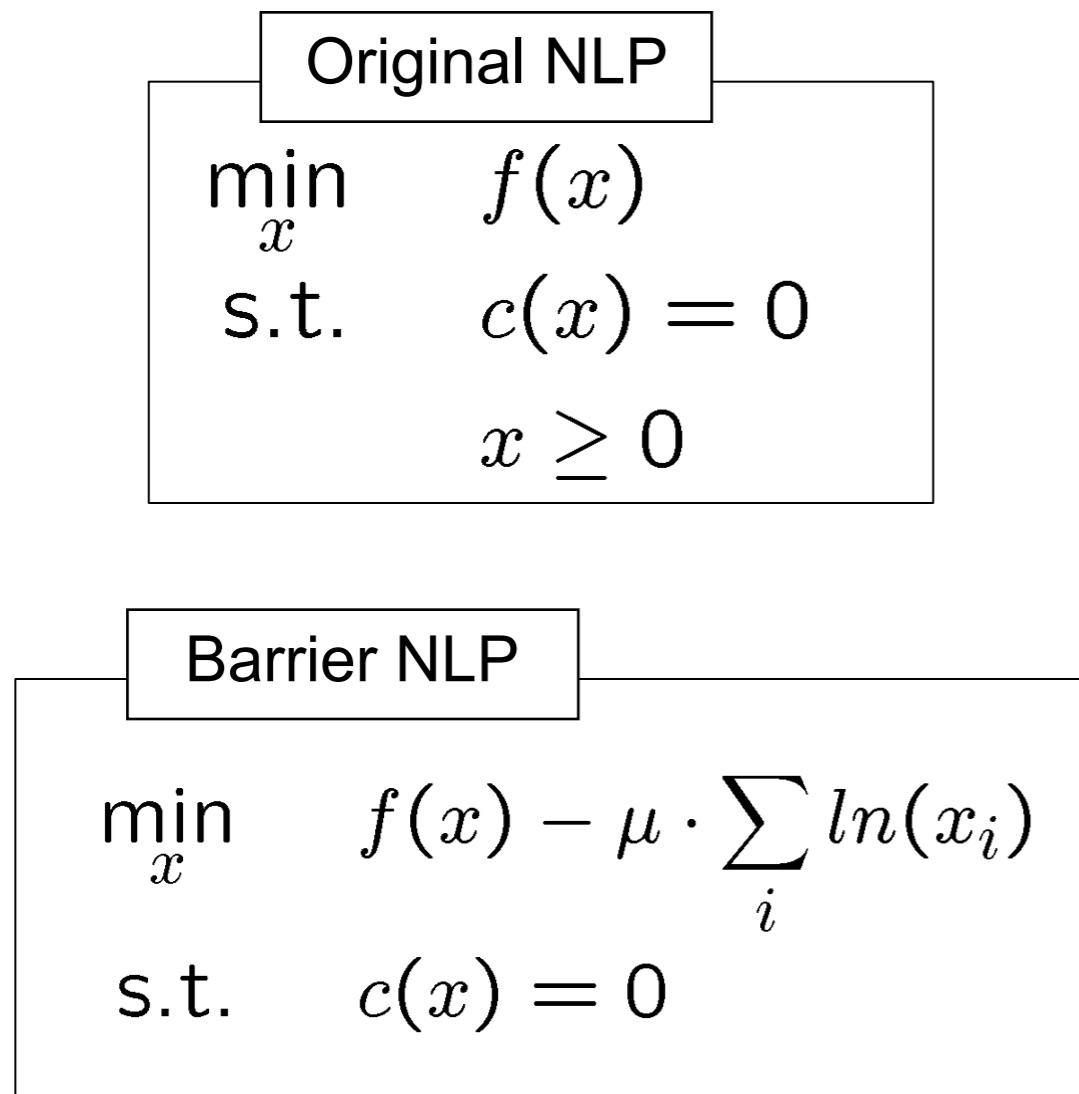


Building the model with Pyomo and PySP

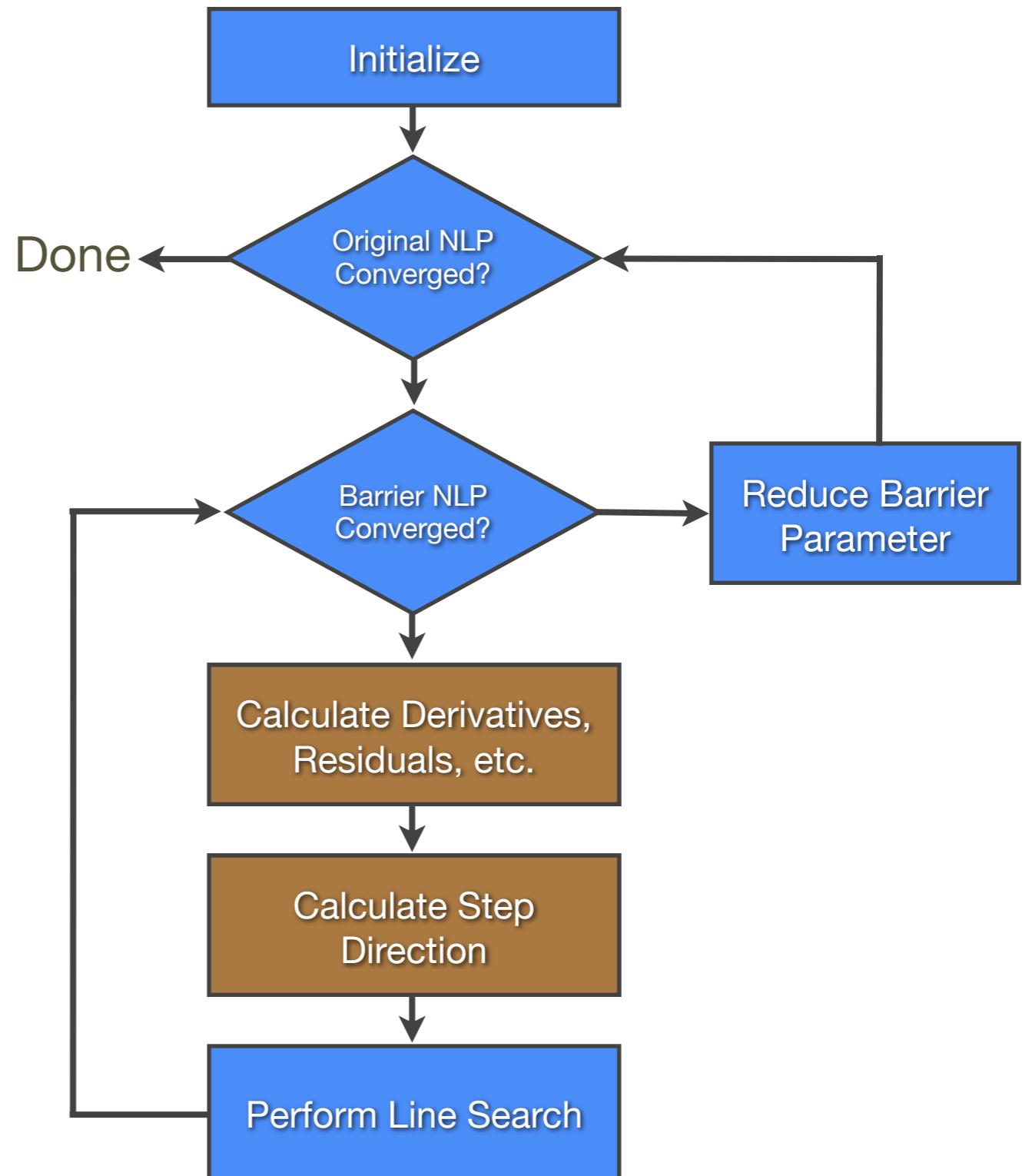


Integrated
Python-based
toolchain

Nonlinear Interior-Point Methods



KNITRO (Byrd, Nocedal, Hribar, Waltz)
LOQO (Benson, Vanderbei, Shanno)
IPOPT (Wachter, Laird, Biegler)



Nonlinear Interior-Point Methods

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

$$\begin{aligned} \min_x \quad & f(x) - \mu \sum_i \ln(x_i) \\ \text{s.t.} \quad & c(x)=0 \end{aligned}$$

$$\begin{aligned} \nabla f(x) + \nabla c(x)^T \lambda - z &= 0 \\ c(x) &= 0 \\ Xz &= \mu e \\ (x > 0, z > 0) \end{aligned}$$

$$z = \mu X^{-1} e$$

$$\begin{aligned} \nabla f(x) + \nabla c(x)^T \lambda - \mu X^{-1} e &= 0 \\ c(x) &= 0 \\ (x > 0) \end{aligned}$$

$$\begin{bmatrix} W_k + \Sigma_k + \delta_w I & \nabla c(x_k) \\ \nabla c(x_k)^T & -\delta_c I \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta \lambda \end{bmatrix} = - \begin{bmatrix} \nabla \psi_\mu(x_k) + \nabla c(x_k)^T \lambda_k \\ c(x_k) \end{bmatrix}$$

$$(W_k = \nabla_{xx}^2 \mathcal{L}, \Sigma_k = Z_k X_k^{-1})$$

Nonlinear Interior-Point Methods

$$\begin{array}{ll} \min_x & f(x) \\ \text{s.t.} & c(x)=0 \\ & x \geq 0 \end{array}$$



$$\begin{array}{ll} \min_x & f(x) - \mu \sum_i \ln(x_i) \\ \text{s.t.} & c(x)=0 \end{array}$$

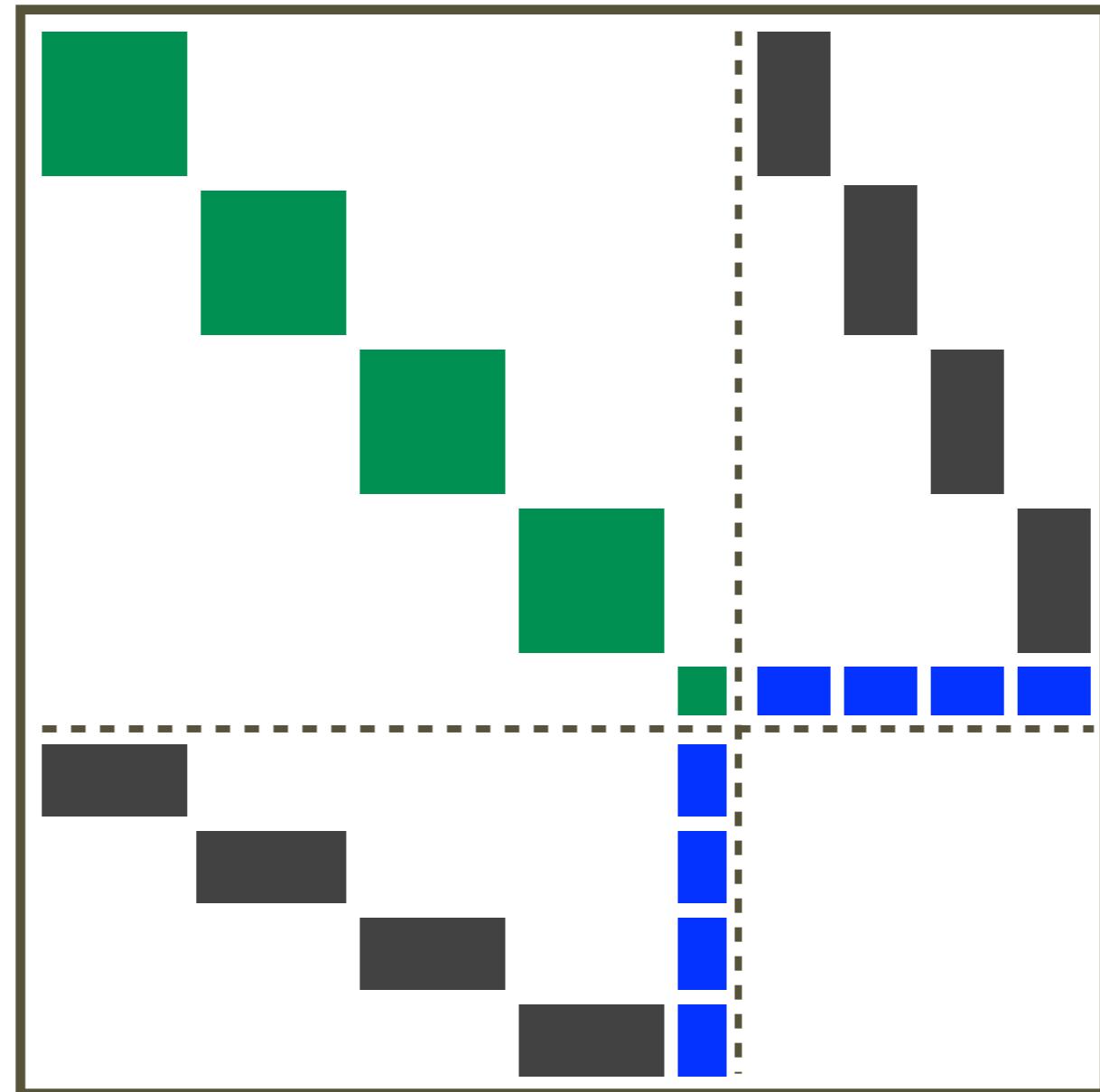
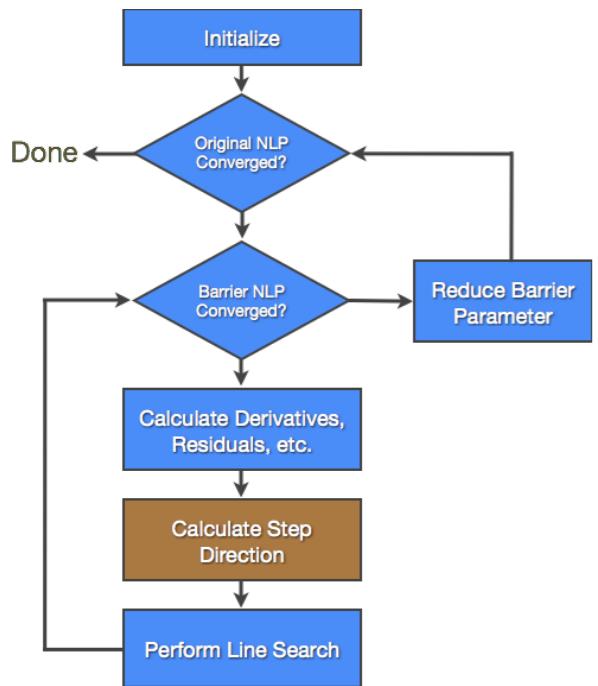
∇f

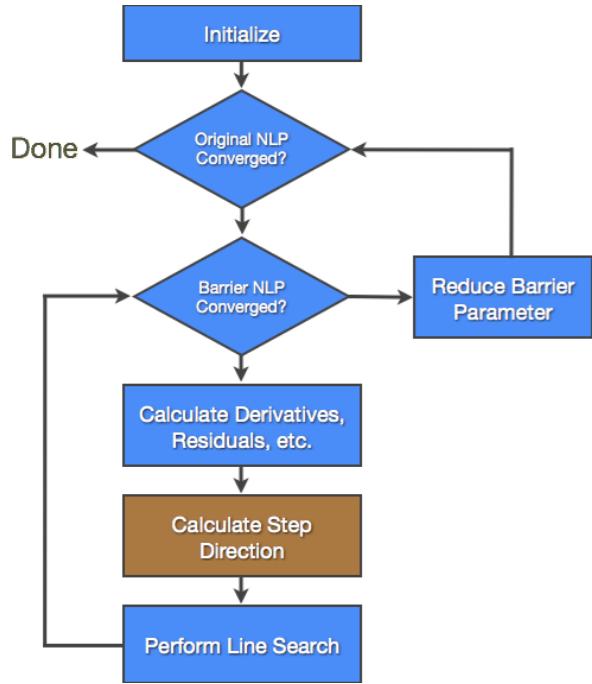
Structure in the optimization problem
Structure in the linear system

$(x > 0, \lambda > 0)$

$$\begin{bmatrix} W_k + \Sigma_k + \delta_w I & \nabla c(x_k) \\ \nabla c(x_k)^T & -\delta_c I \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta \lambda \end{bmatrix} = - \begin{bmatrix} \nabla \psi_\mu(x_k) + \nabla c(x_k)^T \lambda_k \\ c(x_k) \end{bmatrix}$$

$$(W_k = \nabla_{xx}^2 \mathcal{L}, \Sigma_k = Z_k X_k^{-1})$$





$$K_1$$

$$K_2$$

$$K_3$$

$$K_4$$

$$A_1 \\ A_2 \\ A_3 \\ A_4 \\ S$$

$$\left[D_y - \sum_{q \in Q} A_q^T K_q^{-1} A_q \right] \Delta y = r_y - \sum_{q \in Q} A_q^T K_q^{-1} r_q$$

$$\left[D_y - \sum_{q \in Q} A_q^T K_q^{-1} A_q \right] \Delta y = r_y - \sum_{q \in Q} A_q^T K_q^{-1} r_q$$

1: for each i in $1, \dots, n_e$

Factor K
Blocks

1.1: factor K_i

2: let $S = [-\delta_c I]$

3: let $r_{sc} = r_s$

4: for each i in $1, \dots, n_e$

4.1: for each column j in A_i^T

4.1.1: solve the system $K_i q_i^{<j>} = [A_i^T]^{<j>}$

 Form Schur-
 Complement

4.1.2: let $S^{<j>} = S^{<j>} + A_i q_i^{<j>}$

4.2: solve the system $K_i p_i = r_i$

4.3: let $r_{sc} = r_{sc} - A_i p_i$

5: solve $S \Delta \nu_s = r_{sc}$ for $\Delta \nu_s$

Solve Schur-Complement

6: for each i in $1, \dots, n_e$

Solve Remaining
Vars

6.1: solve $K_i \Delta \nu_i = r_i - A_i^T \Delta \nu_s$ for $\Delta \nu_i$

$$\left[D_y - \sum_{q \in Q} A_q^T K_q^{-1} A_q \right] \Delta y = r_y - \sum_{q \in Q} A_q^T K_q^{-1} r_q$$

~~1: for each i in $1, \dots, n_e$~~

Factor K
Blocks

1.1: factor K_i

~~2: let $S = [-\delta_c I]$~~

~~3: let $r_{sc} = r_s$~~

~~4: for each i in $1, \dots, n_e$~~

4.1: for each column j in A_i^T

4.1.1: solve the system $K_i q_i^{<j>} = [A_i^T]^{<j>}$

Form Schur-
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Solve Schur-Complement

~~6: for each i in $1, \dots, n_e$~~

Solve Remaining
Vars

6.1: solve $K_i \Delta \nu_i = r_i - A_i^T \Delta \nu_s$ for $\Delta \nu_i$

Parallel Nonlinear Programming with Schur-complement Decomposition

OOPS [Gondzio & Gothrey, 2007, 2009]

PIPS-NLP [Chiang et al. 2014, Lubin et. al, 2011]

PRBLOCK_IP [Castro 2007]

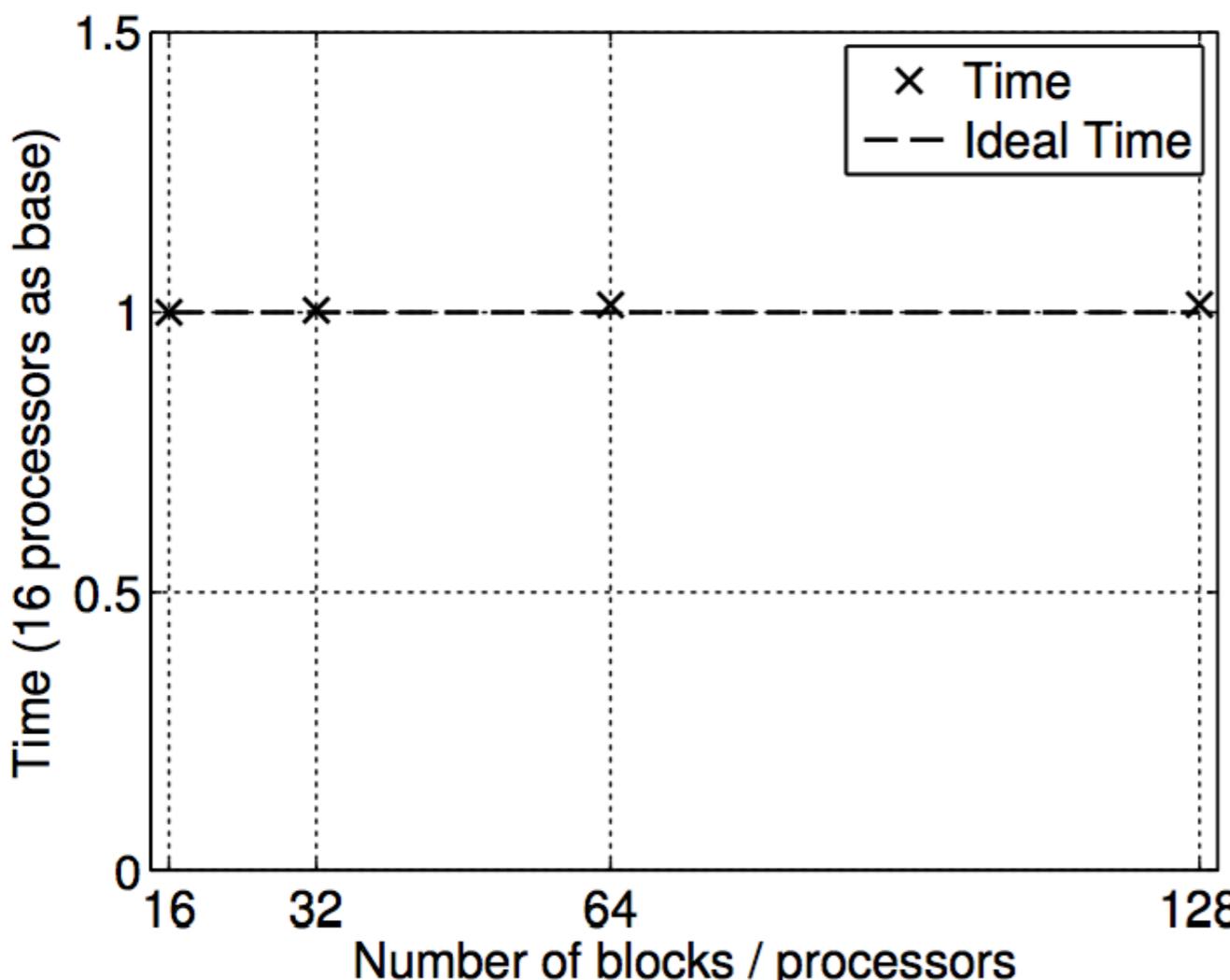
Schur-IPOPT [Kang et al. 2014, Zavala et al., 2008]

- Parameter estimation in LDPE process [Zavala et al. 2008]
- Spatial decomposition of water network [Zhu, 2011]
- Optimal operation of air separation plants with uncertain demand and energy pricing [Zhu, Legg, Laird, 2011a,b]
- Temporal decomposition for dynamic optimization [Word et al. 2014]

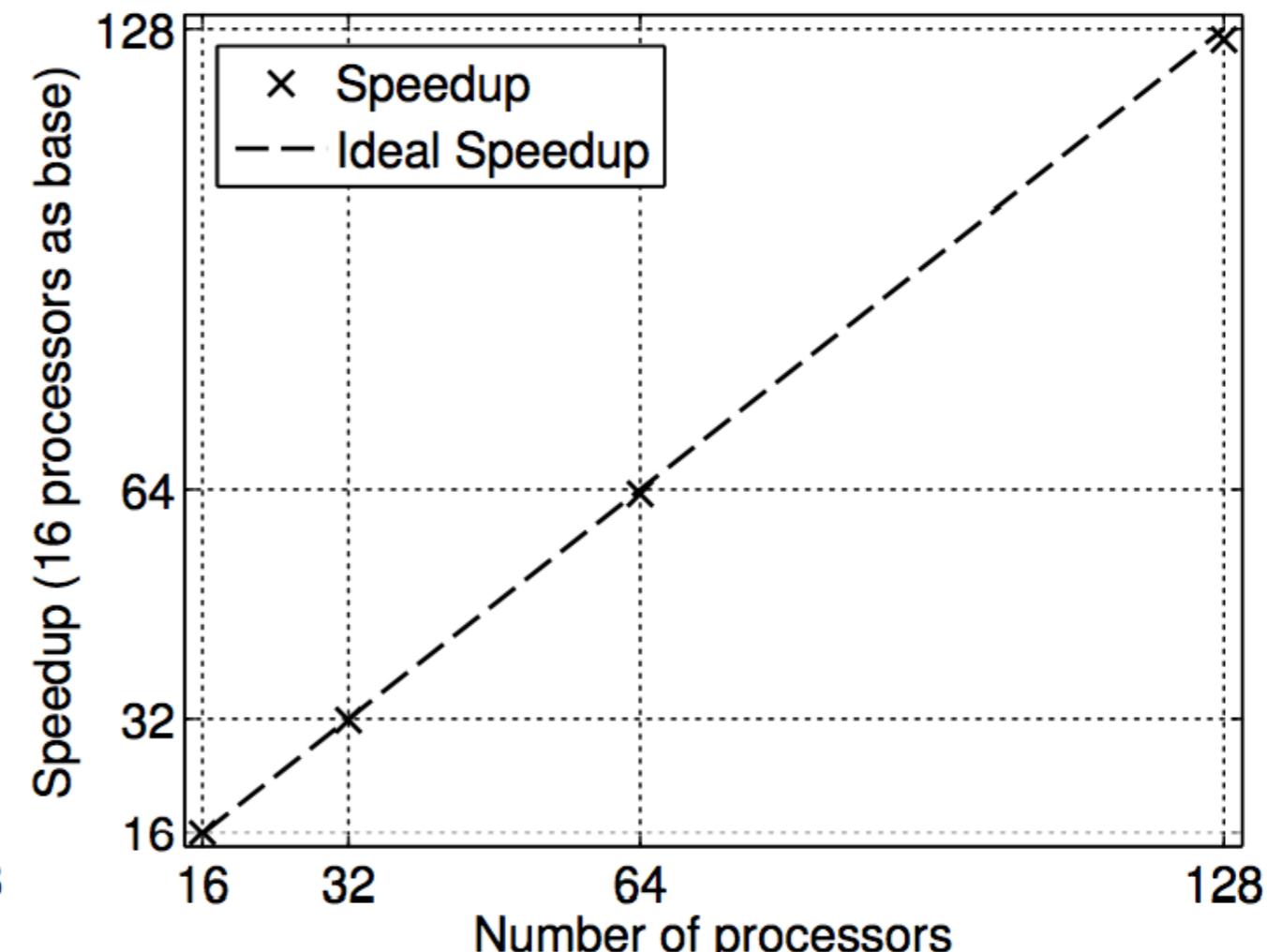
- Implicit Schur-complement approach [Kang et al. 2014]
 - Never explicitly form the Schur-complement
 - Solve it iteratively with PCG using preconditioner from Morales & Nocedal (2000)

Parallel Performance: Dynamic Optimization Under Uncertainty

Dynamic optimization under uncertainty problem
1200 coupling variables - over 15 million total variables



(a) Weak Scaling



(b) Strong Scaling

Kang, J., Word, D.P., and Laird, C.D., "An interior-point method for efficient solution of block-structured NLP problems using an implicit Schur-complement decomposition", to appear in Computers and Chemical Engineering, 2014.

Contingency-constrained ACOPF results

Problem data: case118 distributed with Matpower 4.1

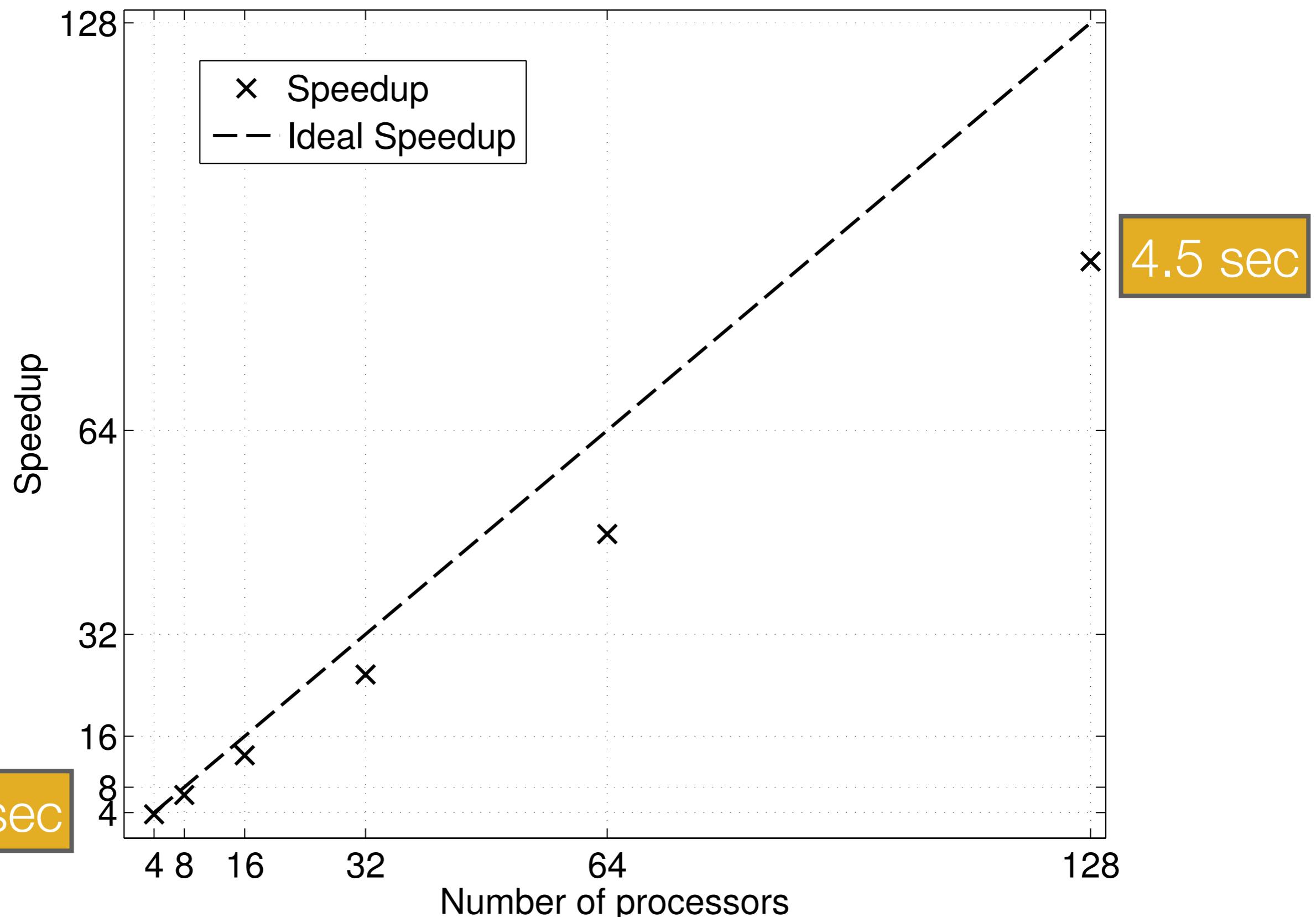
- 118 buses, 54 active generators, and 186 branches

Multi-scenario problem with 128 scenarios in total

- Normal operating scenario and 127 contingencies
- Problem size: ~400,000 variables and ~385,000 constraints

*Wall-clock time obtained from the Red Mesa supercomputing cluster at Sandia National Laboratories. Each node has 12 GB RAM and two, 2.93 GHz quad-core, Nehalem X5570 processors.

Contingency-constrained ACOPF results



Summary and Conclusions

Contingency-constrained ACOPF

- two-stage nonlinear stochastic programming problem
- straightforward implementation within Pyomo/PySP
- parallel IP methods show significant speedup

PySP provides access to other parallel solvers.

Progressive Hedging (PySP, Watson & Woodruff)

- IEEE-57, Parallel solution time: 9 seconds (under 10 iterations).
- case118, Parallel solution time: 20 seconds (under 10 iterations).
- case2383wp, (current work)

Increasing demand for solution of challenging large-scale nonlinear programming problems

- need for parallel algorithms
- opportunities for improving design and operation

Acknowledgements

- Current Students/Researchers

- Jia Kang
- Arpan Seth
- Yankai Cao
- Jianfeng Liu
- Michael Bynum

- Former Students/Postdocs

- Yu Zhu
- Ahmed Rabie
- George Abbott III
- Chen Wang
- Sean Legg
- Daniel Word
- Angelica Wong
- Xiaorui Yu
- Gabriel Hackebel
- Shawn McGee
- Alberto Benavides-Serrano

- Collaborators

- Anya Castillo - Johns Hopkins U. / FERC
- D. Cummings - JHSPH
- S. Iamsirithaworn - MPHT
- W. Hart, S. McKenna, K. Klise, **J.P. Watson**
- C. Silva-Monroy - Sandia**
- T. Haxton, R. Murray - EPA
- Johan Akesson, Lund University
- S. Davis, A. Bratteteig - GexCon
- M. S. Mannan - MKOPSC

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School of Chemical Engineering

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Support

- National Science Foundation Cyber-Enabled Discovery and Innovation (CDI)-Type II
- National Science Foundation (CAREER Grant CBET# 0955205).
- Sandia National Laboratories
- MKO Process Safety Center
- P2SAC