

Stochastic Optimization to Enhance Resiliency and Response Strategies in Critical Infrastructure

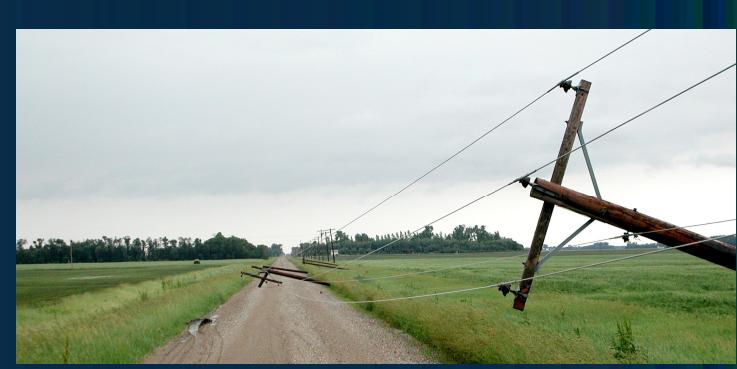
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Problem

Critical infrastructure must be operated robustly and safely under abnormal conditions

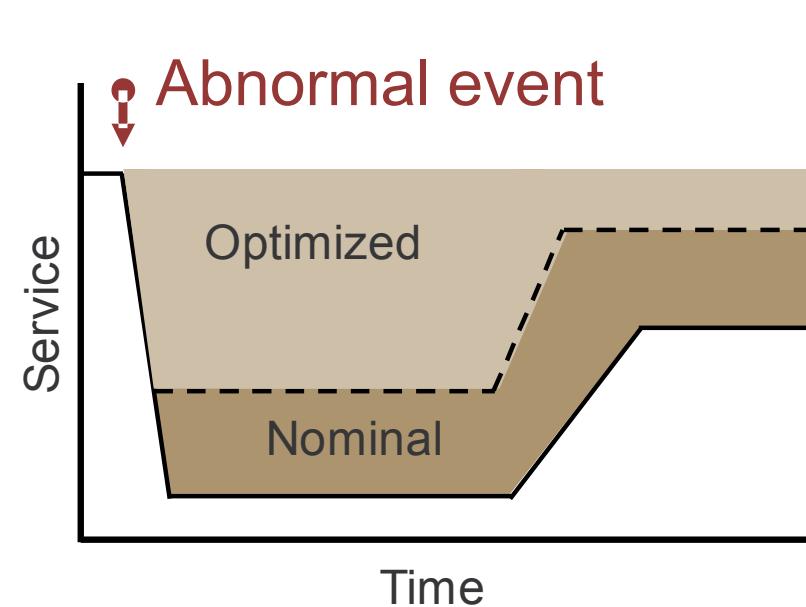
Power systems vulnerable to extreme weather and adversaries



Water systems vulnerable to contamination and loss of service



Petroleum and chemical facilities vulnerable to flammable and toxic gas release



Optimal strategies for hardening and design of monitoring and mitigation systems

Real-time tools for optimal pre-positioning, response, and recovery

Approach

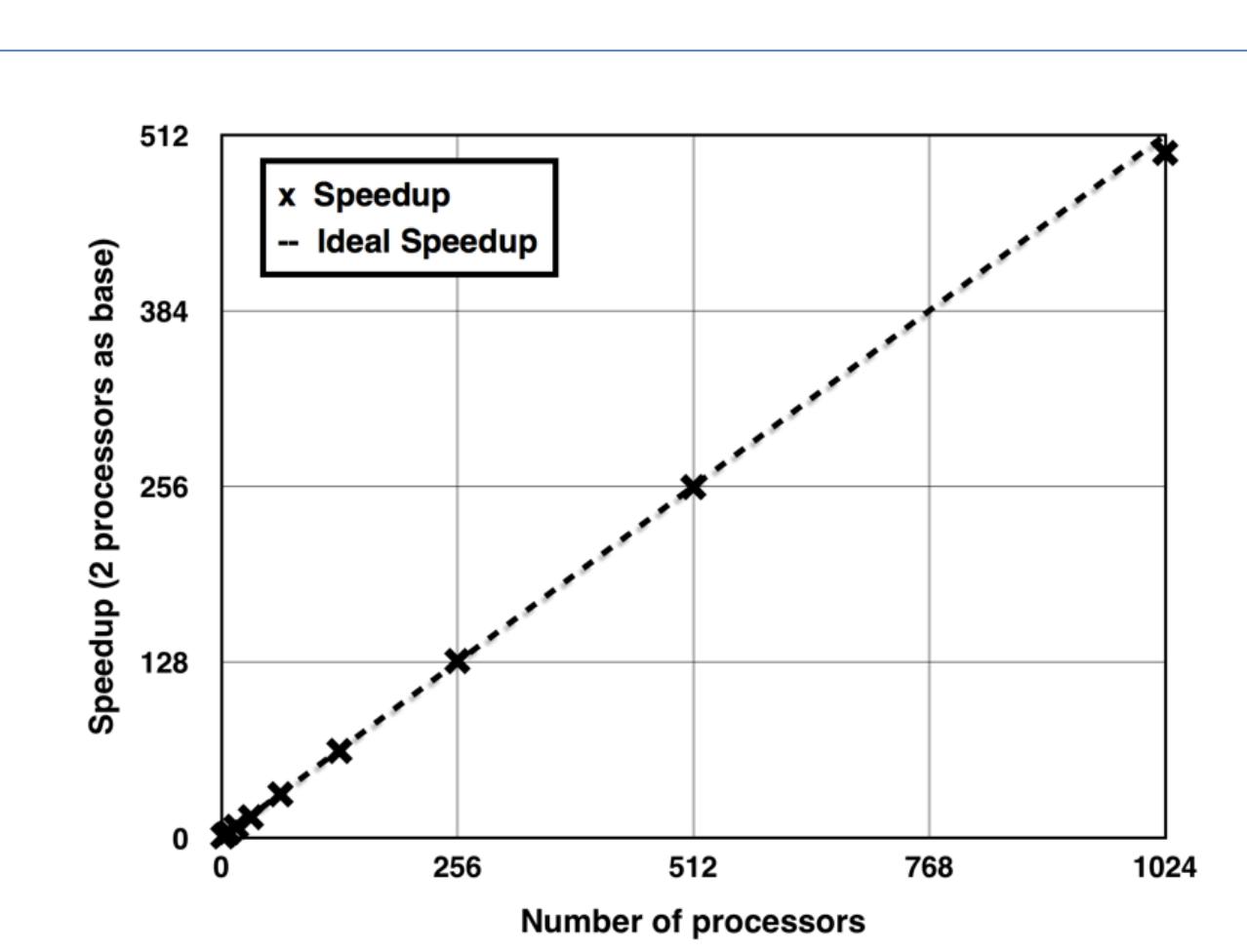
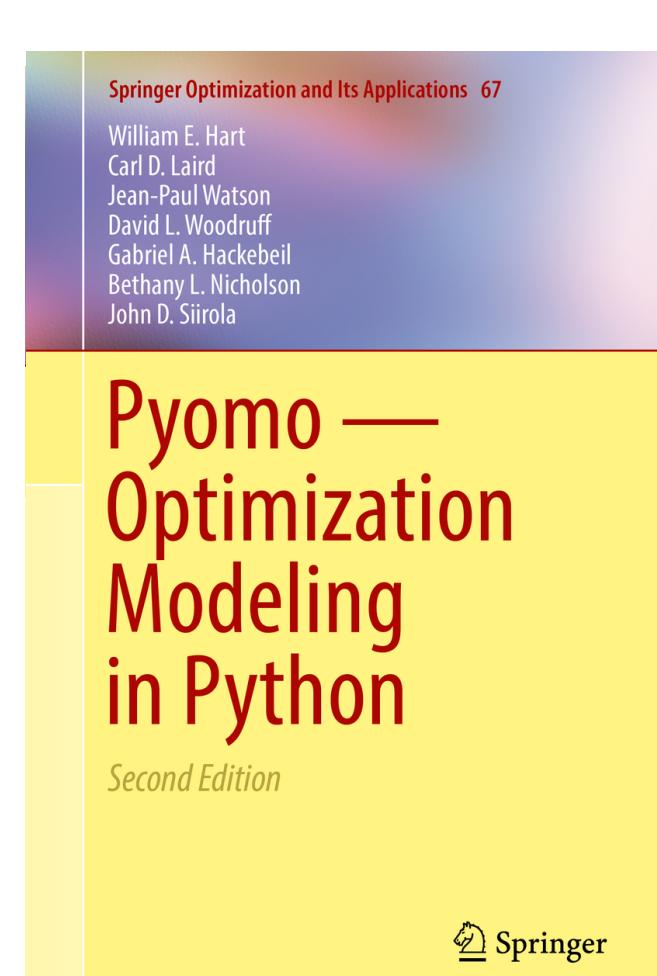
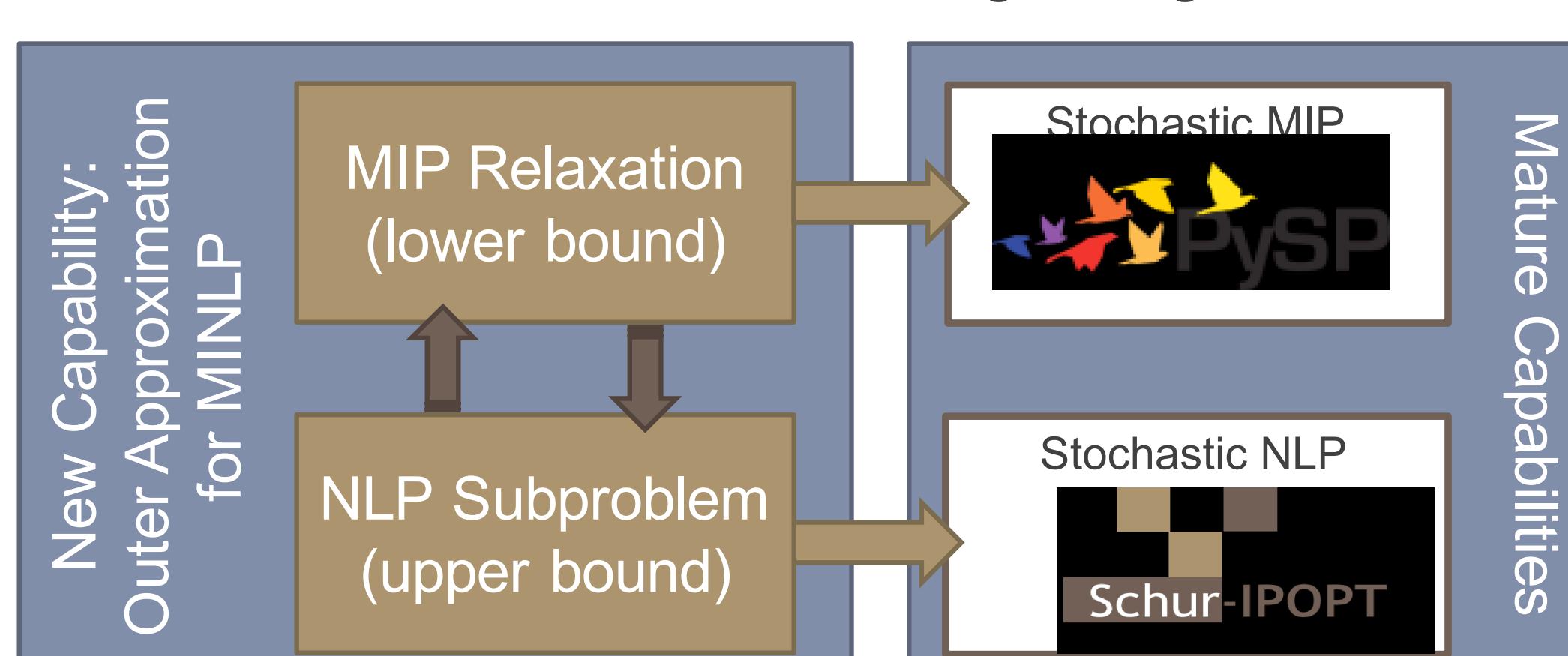
Characteristics: large-scale, uncertain, **discrete & nonlinear**

Existing Capabilities [1-3]:

- stochastic mixed-integer linear programming (MIP)
- stochastic nonlinear programming (NLP)
- parallel decomposition strategies

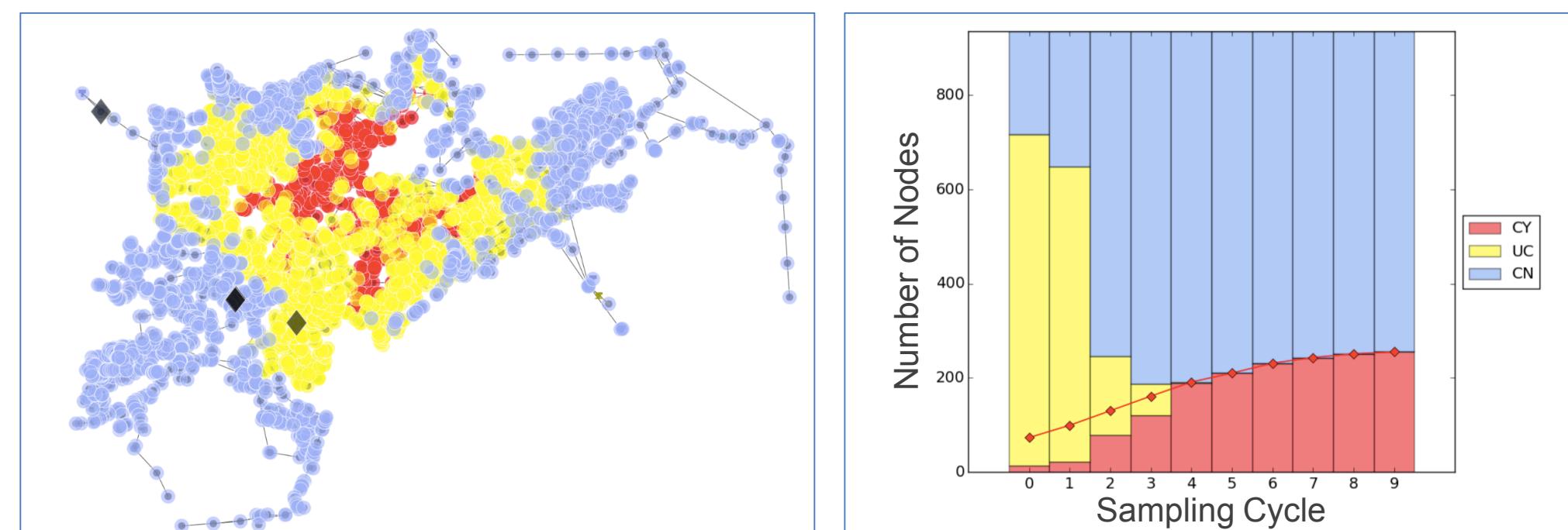
New Capabilities:

MINLP-CRIT: mixed-integer nonlinear programming (MINLP)
EGRET: Electrical Grid Research and Engineering Tool



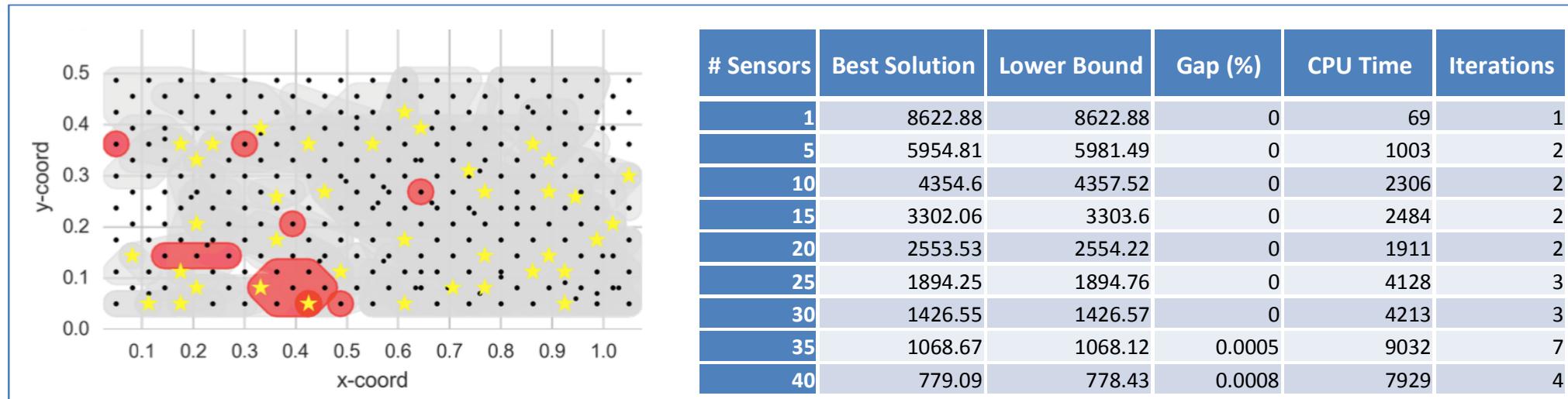
Results

Optimal real-time sampling for plume identification



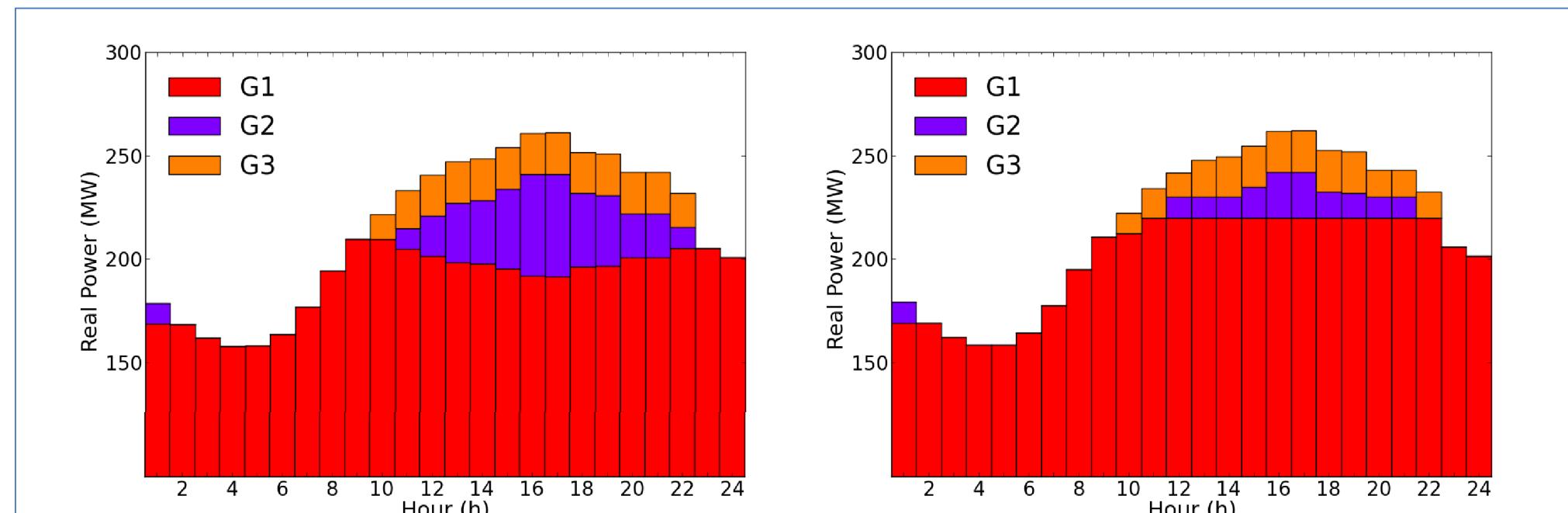
MINLP strategies to determine optimal sampling locations can be coupled with Bayesian statistics to rapidly resolve plume extent with few sampling cycles [4]

Optimal gas detector placement in offshore facilities



Tractable solution of detector placement under non-uniform uncertainty outperforms current industrial practice based on standards recommendations

Optimal Unit Commitment with Nonlinear Physics



Optimal unit commitment with rigorous nonlinear physics yields different generator commitment than linearized model (with significant savings)

Significance

Rigorous discrete optimization with high-fidelity physics

- Nonlinear unit commitment problem (IEEE case 118)
- **Linearized solution infeasible**
- **Global solution \$3.6 M/yr better than local**
- Scalable N-1 Contingency-constrained ACOPF
- **Optimal solution with reliability guarantee**

Optimal sensor placement and sampling strategies

- Contaminant sampling in water distribution systems
 - **Planned application with US EPA for Flint, MI**
- Optimal gas detector placement in process facilities
 - **Approach outperforms existing practice**
 - **Non-uniform sensor failure probability**

[1] Hart, W.E., Laird, C.D., Watson, J.P., Woodruff, D.L., Hackebeil, G.A., Nicholson, B.L., Siirola, J.D. "Pyomo — Optimization Modeling in Python, 2nd Ed.", Springer Optimization and Its Applications, Springer International Publishing AG.

[2] Kang, Jia, Yankai Cao, Daniel P. Word, and Carl D. Laird. "An interior-point method for efficient solution of block-structured NLP problems using an implicit Schur-complement decomposition." Computers & Chemical Engineering 71 (2014): 563-573.

[3] Gade, D., Hackebeil, G., Ryan, S. M., Watson, J. P., Wets, R. J. B., & Woodruff, D. L. (2016). Obtaining lower bounds from the progressive hedging algorithm for stochastic mixed-integer programs. Mathematical Programming, 157(1), 47-67.

[4] Rodriguez, J.S., Bynum, M., Hart, D., Laird, C., Klise, K., and Haxton, T., "Optimal sampling locations to reduce uncertainty in contamination extent in water distribution systems", Submitted to EPA