

A Few Problems to Motivate Numerical Libraries for Grid Computations

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<http://public.lanl.gov/rbent/SmarterGrids/>

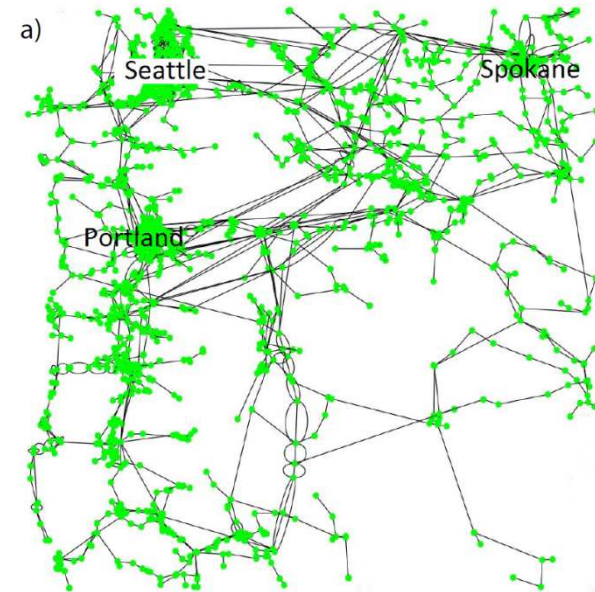
Part 1: Existence a Feasible Steady State in a Grid Subject to Failures—the “N-1 Problem”

Deterministic generation—“N-1” security

1. Remove a major component (~30,000)
2. Compute new power flow solution (~ 1 msec)
3. Verify there are no overloaded lines or generators or voltages out of bounds
4. Estimate of total time ~ 30 seconds
5. Embarrassingly parallel problem

Deterministic generation—“N-2” security

1. Remove two major components (~30,000 X 30,000)
2. 30 seconds → 250 hours
3. Even embarrassingly parallel won't help



“N-1” works and “N-2” is typically not required.....because component failures are rare and the probability of two occurring (“N-2”) with sufficient spatiotemporal correlation to interact and cause problems is exceedingly rare.....assuming failures are independent....

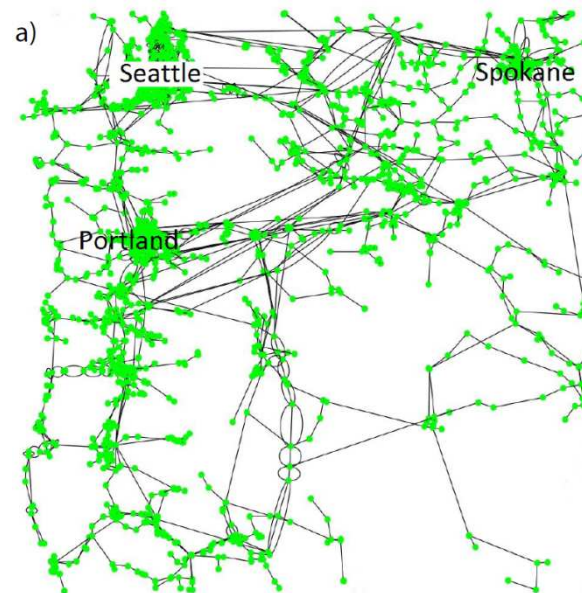
Existence of a Feasible Steady State—Add 100 Wind/Stochastic Generators to the Grid—the “N-a Problem”

Concept #1—“N-1” checks for each wind generator

1. “Fail” wind generators one at a time
2. Only adds 100 cases to the ~30,000

Why doesn’t this work?

1. Wind generator “failure” is far more probable than component failure
2. Multiple generators will “fail” at the same time→Need to check (all possible configurations of failure) X (all N-1 cases).
 $\sim 2^{100} \times 30,000 \sim 10^{24}$ years
3. And it is worse that this... ramps from $\frac{1}{2}$ output to full output can be just as problematic



How many states do we assign to each stochastic generator? 3, 4....10?

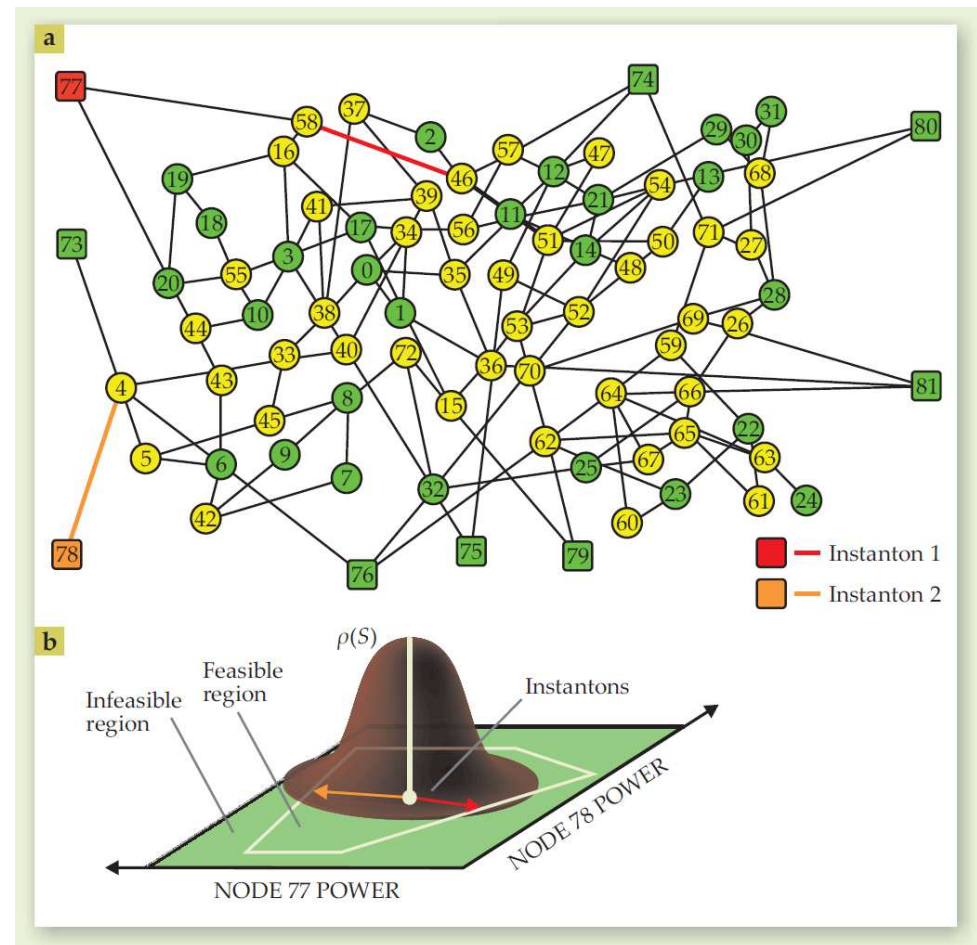
This approach does not scale.....

Better search methods are needed to find the most probably (yet rare) modes of failure

Existence of a Feasible Steady State With Stochastic Generation—A Fundamentally New Approach Is Needed

One Illustrative Example—that leverages extreme statistics methods from statistical mechanics

1. Consider full probability distributions over stochastic generation
2. Form the feasible space by taking the intersection of all N network constraints—turns out to be a convex region
3. Find the locations of highest probability of the stochastic generation on the boundary of the feasible region.
4. If feasible region is convex.... #3 is found by solving N linear programs.
5. Computational complexity reduced from say 2^{100} power flows to 100 LP's



Probabilities of Infeasible Steady State Can Be Built Into an OPF—Chance Constrained OPF

Demonstrated computational tractability on a large network— ~4000 lines and ~400 generators—Polish grid from Matpower

R. Bent, D. Bienstock, and M. Chertkov, Synchronization-Aware and Algorithm-Efficient Chance Constrained Optimal Power Flow, proceedings of 2013 IREP Symposium-Bulk Power System Dynamics and Control, [arxiv:1306.2972](https://arxiv.org/abs/1306.2972)

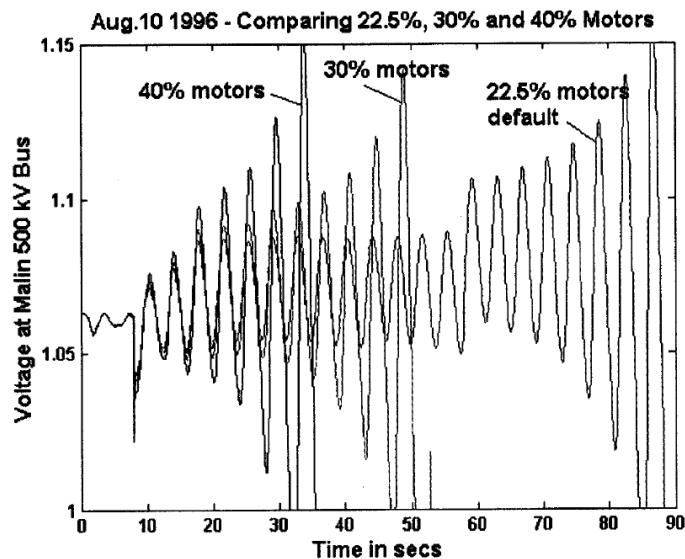
$$\begin{aligned} \min_{p, \alpha} \mathbb{E}_{\mathbf{w}} [f(p - (e^T \mathbf{w})\alpha)] \\ \text{s.t. } \sum_{i \in G} \alpha_i = 1, \quad \alpha \geq 0, \quad p \geq 0 \\ \text{Prob (PF Eqs. are not feasible)} < \varepsilon \\ \text{Prob} (\beta_{ij} |\sin(\theta_i - \theta_j)| > \bar{p}_{ij}) < \epsilon_{ij} \quad \forall_{i,j \in \mathcal{E}} \\ \text{Prob} (p_g - (e^T \mathbf{w})\alpha_i > p_g^{max}) < \epsilon_g \quad \forall_{g \in G} \\ \text{Prob} (p_g - (e^T \mathbf{w})\alpha_i < p_g^{min}) < \epsilon_g \quad \forall_{g \in G} \end{aligned}$$

Numerical libraries needed to enable this family of approaches: ????

Part 2: Linear and Nonlinear Dynamical Models of Aggregated Distribution Loads (and Generators...)

Linear models of aggregated load

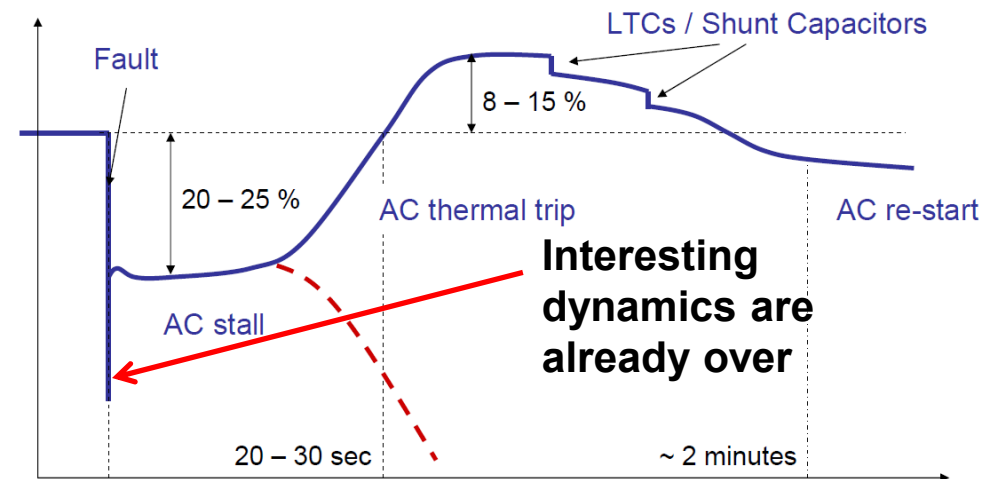
Uncertainty/Inaccuracy of linear dynamical load models creating uncertainty in estimates of small signal stability.



Pereira, L.; Kosterev, D.N.; Mackin, P.; Davies, D.; Undrill, J.; Wenchun Zhu, "An interim dynamic induction motor model for stability studies in the WSCC," *Power Engineering Society General Meeting, 2003, IEEE*, vol.3, no., pp.1404,1404, 13-17 July 2003

Nonlinear models of aggregated load

Close in fault causes induction motor stalling...subsequent large reactive currents keep voltage depressed and motor stalled.



Fault-Induced Delayed Voltage Recovery

Nonlinear Dynamical Models of Aggregated Distribution Loads

Power flow equations

$$\partial_z \rho = -p - r \frac{\rho^2 + \phi^2}{v^2},$$

$$\partial_z \phi = -q - x \frac{\rho^2 + \phi^2}{v^2},$$

$$\partial_z v = -\frac{r\rho + x\phi}{v}.$$

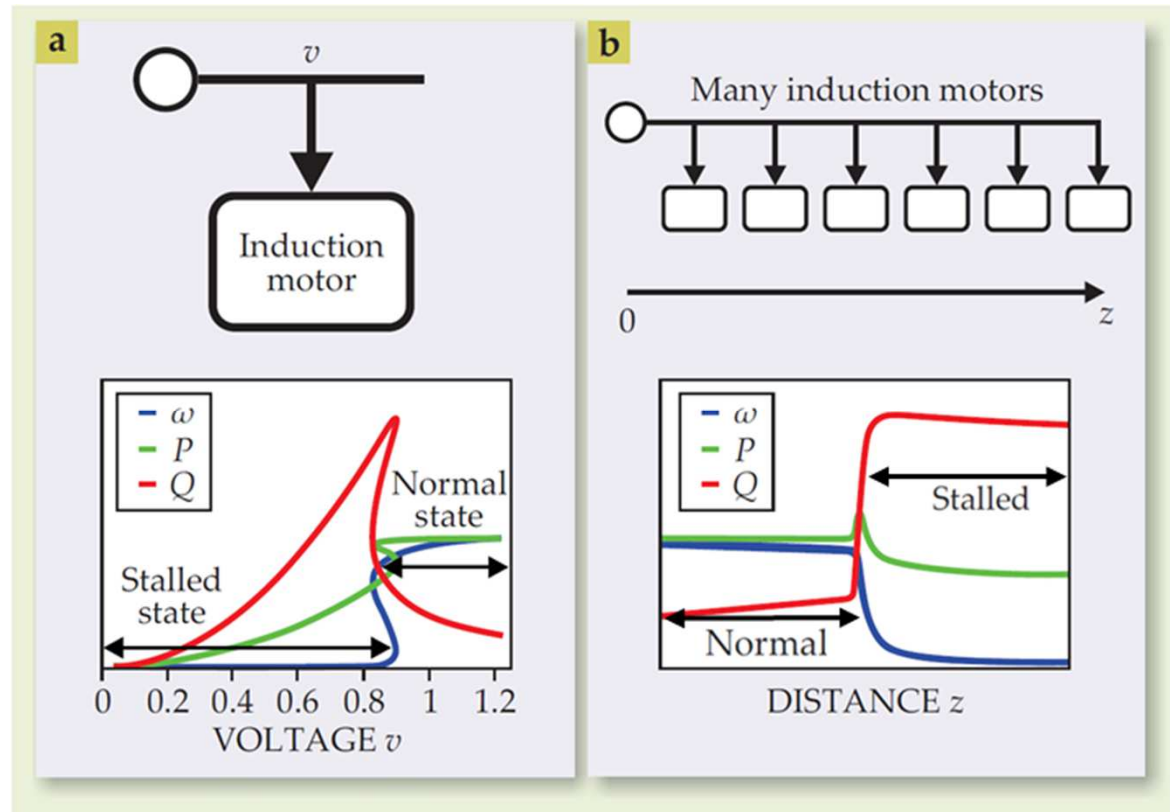
Motor dynamics

$$\mu \frac{d}{dt} \omega = \frac{p}{\omega_0} - t_0 \left(\frac{\omega}{\omega_0} \right)^\alpha,$$

$$p = \frac{s r_m v^2}{r_m^2 + s^2 x_m^2},$$

$$q = \frac{s^2 x_m}{r_m^2 + s^2 x_m^2} v^2.$$

Coupling of power flow and motor dynamics creates a PDE that demonstrates collective behavior/collapse



Duclut, Charlie and Backhaus, Scott and Chertkov, Michael. Hysteresis, phase transitions, and dangerous transients in electrical power distribution systems. Phys. Rev. E **87**, 2013

Nonlinear Dynamical Models of Aggregated Distribution Loads

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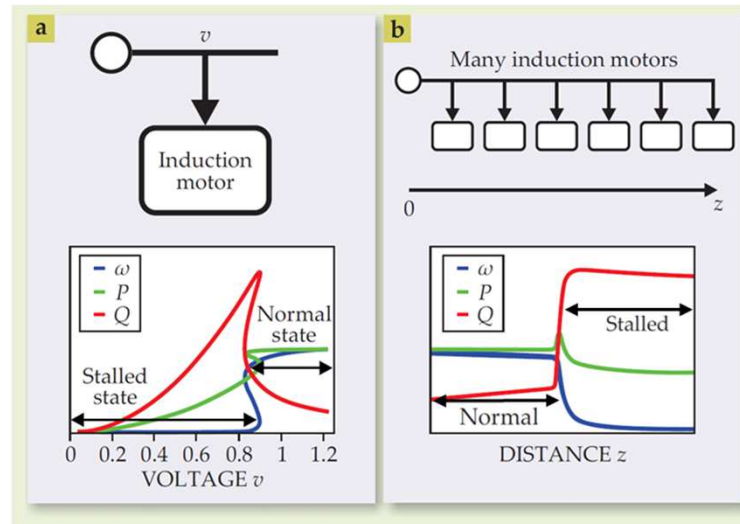
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1. PDEs provide a good description of the dynamics
2. Adding a PDE at each feeder immediately leads to computational intractability
3. Need to reduce the PDE to several ODE's at each feeder to even come close to a tractable problem
4. Model reduction of the nonlinear behavior is difficult

Numerical library needs—generalized linear and nonlinear equivalent models



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