

Stochastic Modeling and Supercomputing for Smart Grids

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Smart Grids and Supercomputing

- Smart grids will generate large amounts of sensor measurements whose timely processing and utilization will be used to improve efficiency and reliability
- Intermittent energy sources (wind, solar) increase supply uncertainties
- Parallel computing provides natural tools for data processing, optimal decision making and uncertainty modeling



Advantages of HPC

- High performance compilers
- Parallelism: doing many things at the same time
 - Instruction-level parallelism: doing multiple operations at the same time within a single processor (e.g., add, multiply, load and store simultaneously)
 - Multiprocessing: multiple CPUs working on different parts of a problem at the same time
 - Shared Memory Multithreading
 - Distributed Multiprocessing
- Powerful storage hierarchy



What is High Performance Computing?

- High Performance Computing (HPC), also called supercomputing, is the biggest, fastest computing right this minute. Likewise, a supercomputer is one of the biggest, fastest computers right this minute. So the definition of supercomputing is constantly changing.
- New York Blue:
 - Consists 18 racks IBM Blue Gene
 - Each rack consists of 1024 compute nodes (a total of 18432 nodes)
 - each node consists two 700 MHz PowerPC 440 core processors and 1 GB of memory (a total of 36864 processors and 18.4 TB of memory)
 - Website: <http://www.bnl.gov/newyorkblue/>
 - Top 500 Supercomputer ranking: 67 as of 06/2010 (17 as of 06/2008)



Some Applications of HPC

- State Estimation
 - Estimate the steady states condition of EPS using online measured values
 - The measurement system consists of active and reactive line power flow and bus injection real and reactive power measurement and bus voltage magnitude measurement
- Forecasting
 - Weather forecasting
 - Load, price, capacity, equipment states, rating and reliability, etc.
- Control and Planning
 - Unit Commitment Problem, Economic Dispatch
- Power Flow Control



State Estimation

- Provide reliable estimates of the quantities required for monitoring and control of the EPS
- A set of measurements obtained is centrally processed by a static state estimator
 - Higher frequency -- shorten the time interval between consecutive state estimations to allow a closer monitoring of the system evolution particularly in emergency situations in which the system state changes rapidly
 - Larger size -- enlarge the supervised network by extending state estimation to low voltage sub networks



State Estimation

- Challenges:

- Higher frequency requires the development of faster state estimation algorithms
- Larger size increase the demand on the numerical stability of the algorithms

- Solutions:

- Parallel and distributed implementations of the state estimation function



State Estimation

- State Estimation model:

$$\mathbf{z} = \mathbf{h}(\mathbf{x}) + \mathbf{w}$$

\mathbf{z} – ($m \times 1$) measurement vector

\mathbf{x} – ($2n \times 1$) true state vector

$\mathbf{h}(\cdot)$ – ($m \times 1$) vector of nonlinear functions

\mathbf{w} – ($m \times 1$) measurement error vector

m – number of measurements

n – number of buses



Load Forecasting

- Accurate forecast of future demand required by all entities involved in the energy markets
 - Electric Utilities
 - Independent System Operators
 - Power Marketers
- Different forecast horizons
 - Long Term: Several years out – required for planning purposes
 - Mid Term: Several weeks to months – scheduling maintenance, planning fuel supply, transactions
 - Short Term: Next hour to next week – daily operation, energy transactions, reliability studies

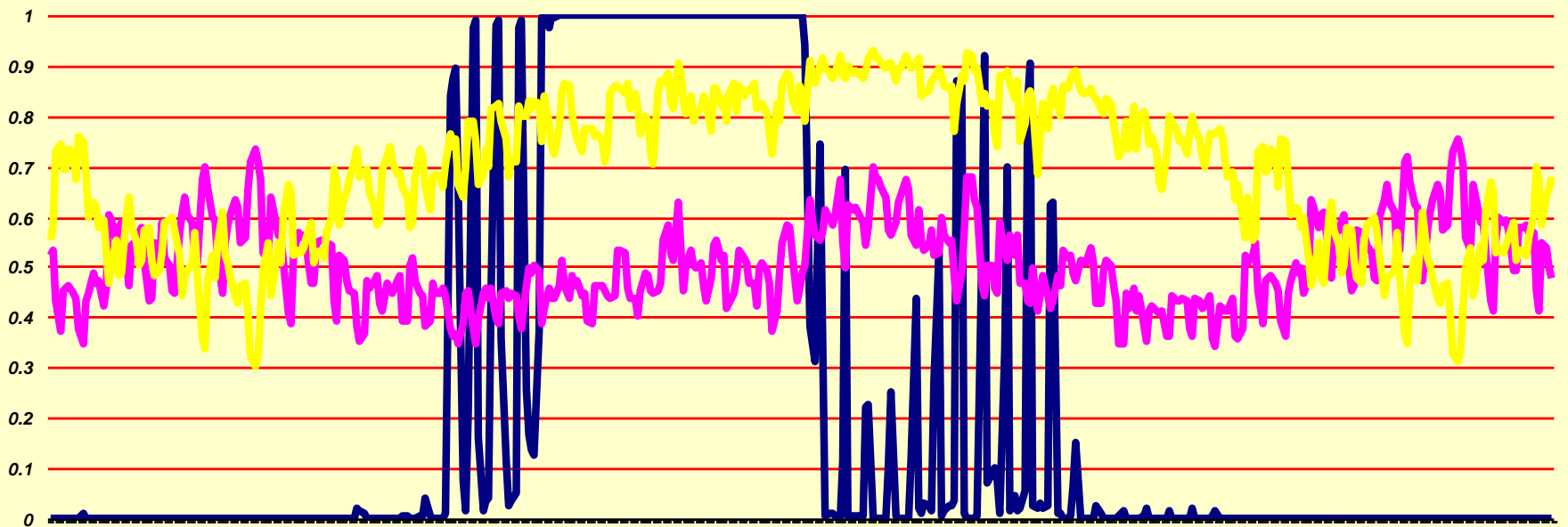


Regression Models: Example

- $L(t) = F(d(t),h(t))*f(w(t))+R(t)$
 - $L(t)$ – Actual load at time t
 - $d(t)$ – day of the week
 - $H(t)$ – hour of the day
 - $F(d,h)$ – daily and hourly component
 - $w(t)$ – weather data that include the temperature and humidity
 - $f(w)$ – weather factor
 - $R(t)$ – random error

Forecasting using ANNs: Example

ANNs Sigmoid Nodes' Output
Seperation of Seasons



$$\text{Predict} = \sum \text{Sigmoid Node Output} + \text{Error}$$



Unit Commitment Problem

- Optimal generator assignment problem for electric grid
- Schedule the power generator units over a short time period, in order to:
 - Minimize the operation cost
 - Satisfy the electricity demand
 - Maintain system reliability
- Formulation: Mixed Integer Programming Problem



Unit Commitment Problem

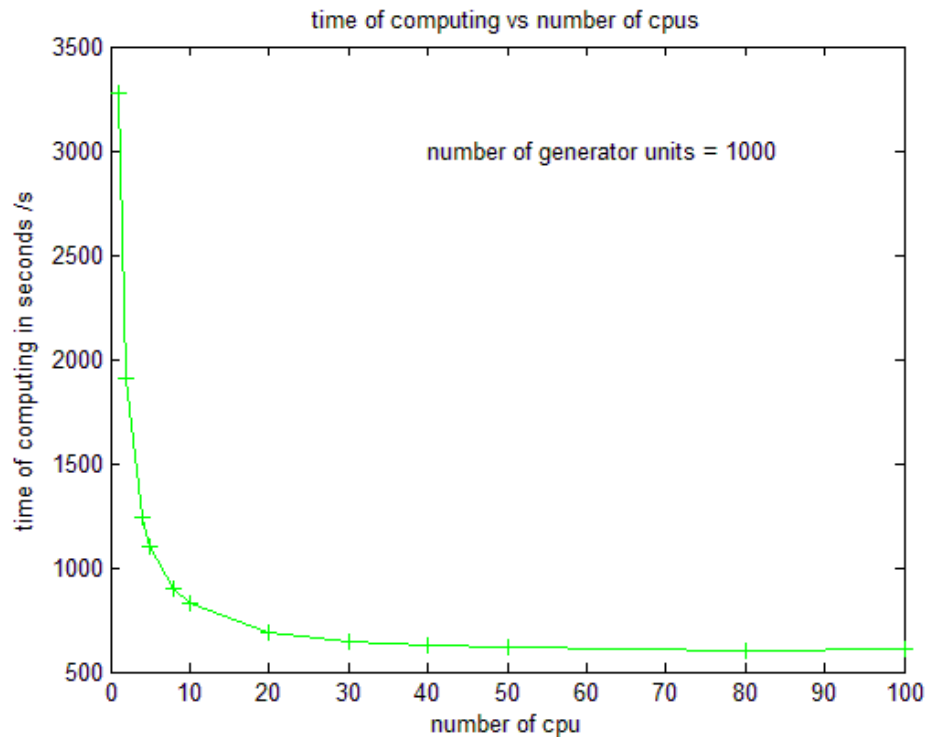
■ Challenges

- Large scale mixed integer programming: thousands of integer and continuous variables, numerous security constraints
- Uncertainty in electricity loads
- Volatile energy sources: wind energy, solar energy

■ Solutions

- Parallel implementation of UCP solver to boost computational speed
- Stochastic modeling of uncertainty

Effectiveness of Parallel Computing



Computational time (in seconds) *vs.* Number of CPU's
Number of power generators: 1000