

#### Accelerating interior point methods with GPUs for smart grid systems

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# Accelerating interior point methods with GPUs for smart grid systems

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## Outline

#### 1 Introduction

- 2 Smart grid test case
- 3 Interior point method

4 Future plans



#### Introduction

- Ph.D. student at GPUlab, DTU IMM since November 2010.
  - Title: Scientific GPU Computing for Dynamical Optimization
  - Investigate GPUs for solving optimization problems, primarily model predictive control (MPC).
  - Initial test problem is distribution and control of electrical power via smart grid through the use of MPC.
- Completed M.Sc. at DTU IMM in October 2010:
  - Title: Implementation and evaluation of fast computational methods for high-resolution ODF problems on multi-core and many-core systems
  - Used GPUs for ray tracing and matrix-free SpMV.

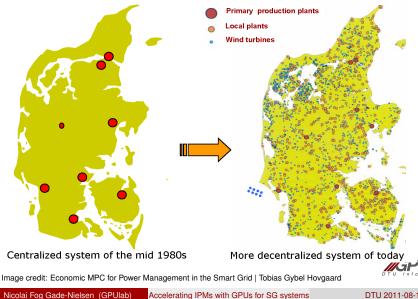


# What is a smart grid?

- Dynamically control energy production and consumption according to some objective, eg:
  - Increased use of uncontrollable renewable power such as wind and solar.
  - Lower production costs.
  - Lower CO2 production.
  - etc.
- Examples
  - Heat pump control in houses
  - Power plant production control



#### Power plants in Denmark

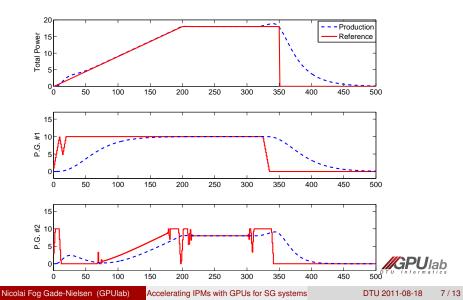


# Simple Economic MPC 1

- Prediction horizon of *N* time steps.
- **\square**  $N_p$  number of power plants with different properties.
- Properties on each power plant:
  - Response time.
  - Minimum and maximum change in one time step.
  - Minimum and maximum power production.
  - Cost.
- Simple example: Two power plants.
  - A cheap but slow power plant.
  - An expensive but fast power plant.



# Simple Economic MPC example



# Simple Economic MPC 2

Define power plant production as an economic MPC problem:

$$\begin{array}{ll} \min & \phi = g's + c'x \\ s.t. & Ax - s = b \\ s \geq 0 \end{array}$$

- c is the cost and x is the control variables of each power plant in each time step.
- $\blacksquare$  s is a slack variable and g is the cost of using it.
- A is our constraint matrix which encodes the constraints of each power plant and dynamics of the system. Very sparse and highly structured.
- **Number of control variables:**  $N_p * N + N$



# IPM Algorithm (without Predictor-Corrector)

- While not converged do
  - Compute duality gap.
  - Set centering parameter.
  - Compute residuals.
  - Solve newton step.

$$\begin{bmatrix} 0 & -A' & 0 \\ -A & 0 & I \\ 0 & S & \Lambda \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta \lambda \\ \Delta s \end{bmatrix} = -\begin{bmatrix} r_L \\ r_s \\ r_\lambda \end{bmatrix}$$
(1)
$$\begin{bmatrix} 0 & -A' \\ -A & -D^{-1} \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta \lambda \end{bmatrix} = -\begin{bmatrix} r_L \\ r_s - \Lambda^{-1} r_\lambda \end{bmatrix}$$
(2)
$$D = S^{-1} \Lambda$$
$$S = diag(s), \Lambda = diag(\lambda)$$

Compute step length.Update step.



### Solving the Newton step

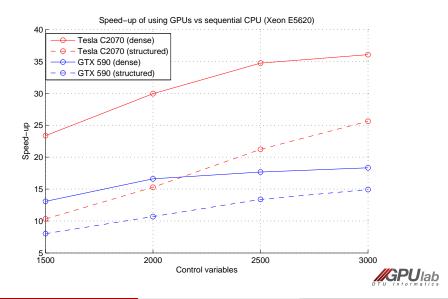
- Bottleneck: Solving the Newton step
- Normally solved with direct method:
  - Compute the Hessian matrix and use Cholesky factorization.

 $H_A = A^T(D)A$ Cholesky :  $H_A = LL^T$ 

- Implemented using CUBLAS and MAGMA.
- What is MAGMA?
  - The MAGMA project aims to develop a dense linear algebra library similar to LAPACK but for heterogeneous/hybrid architectures, starting with current "Multicore+GPU" systems.'
  - Very fast Cholesky factorization.
  - Developed at University of Tennessee.



#### **Results**



#### Problems

Ill-conditioning

Normal equations -> condition number squared

(

$$H_A = A^T(D)A$$
  
Cholesky :  $H_A = LL^T$ 

In the later iterations of the IPM, either  $s_i$  or  $\lambda_i$  goes toward 0.

$$\begin{bmatrix} 0 & -A' & 0 \\ -A & 0 & I \\ 0 & S & \Lambda \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta \lambda \\ \Delta s \end{bmatrix} = -\begin{bmatrix} r_L \\ r_s \\ r_\lambda \end{bmatrix}$$
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(2)  
$$D = S^{-1}\Lambda$$
$$S = diag(s), \Lambda = diag(\lambda)$$

Loss of sparsity

- $N_p = 1000, N = 48$ : 48048 decision variables.
- 48048<sup>2</sup> hessian matrix to factorize: about 17 GB.

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Accelerating IPMs with GPUs for SG systems



#### Future plans

- Iterative methods and preconditioning
  - Identify iterative solvers for solving the newton step iteratively.
  - Identify good preconditioners.
  - Evaluate which of the solvers and preconditioners are suitable for GPU implementation.
- Create toolbox for solving optimization problems using GPUs
- Smart grid problems used as test case for toolbox.

