

#### Block algebraic methods for CT and their performance

 $f(x + \Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f(x + \Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f(x$ 

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#### **DTU Compute**

Department of Applied Mathematics and Computer Science

### Contributions from students etc.

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#### MULTICORE PERFORMANCE OF BLOCK ALGEBRAIC ITERATIVE RECONSTRUCTION METHODS\*

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Abstract. Algebraic iterative methods are routinely used for solving the ill-posed sparse linear



#### The Walnut case













### Forward projection





#### Several different types:

Line, splatting, footprint, interpolation, ext. rays

#### Correction





AA<sup>T</sup> can be pre-calculated (as forward projection).

# **Back projection** Matrix operation: $\mathbf{x} := \mathbf{x} + \lambda \mathbf{A}^{\mathrm{T}} \mathbf{y}$

#### Several different types:

Line, splatting, footprint, interpolation, ext. rays





























#### • Correction $(y = (b-Ax) ./diag(AA^T))$







#### Back projection ( $x := x + \lambda A^T y$ )







#### Back projection ( $x := x + \lambda A^T y$ )



#### SIRT reconstruction in ASTRA



% Set up a GPU reconstruction cfg = astra\_struct('SIRT3D\_CUDA'); cfg.ReconstructionDataId = rec\_id; cfg.ProjectionDataId = proj\_id; cfg.option.GPUindex = 0; % Create the algorithm object alg\_id = astra\_mex\_algorithm('create', cfg); % Run 64 SIRT iterations astra\_mex\_algorithm('iterate', alg\_id, 64); % Get the result rec = astra\_mex\_data3d('get', rec\_id);



#### **Block Methods**

• Forward projection  $(A_0x)$ 



• Correction  $(y = (b - A_0 x) ./diag(A_0 A_0^T))$ 



Back projection ( $x := x + \lambda A_0^T y$ )



• Forward projection  $(A_1x)$ 



• Correction  $(y = (b - A_1 x) ./diag(A_1 A_1^T))$ 



Back projection ( $x := x + \lambda A_1^T y$ )



Forward projection (A<sub>2</sub>x)



• Correction  $(y = (b-A_2x) ./diag(A_2A_2^T))$ 



Back projection ( $x := x + \lambda A_2^T y$ )



Forward projection (A<sub>3</sub>x)



• Correction  $(y = (b-A_3x) ./diag(A_3A_3^T))$ 



Back projection ( $x := x + \lambda A_3^T y$ )



#### Block method using ASTRA



```
for iter = 1:num iter
    for blk = 1:num blks
        % Create the algorithm object
        alg id = astra mex algorithm('create', cfg{blk});
        % Run 1 SIRT iteration with this block
        astra mex algorithm('iterate', alg id, 1);
        % Delete algorithm object to conserve memory
        astra mex algorithm('delete', alg id);
    end
end
% Get the result
rec = astra mex data3d('get', rec id);
```







NTII





#### Block methods converge faster



### Fully optimized block code





GPU	GTX 680	K20	K40	K80(x1)	TITAN X
Runtime (s)	47.0	25.1	20.7	22.8	10.5

#### <1 min is ok! - but what about larger problems?</p>



### Distributed / Multi-GPU computing

#### **Domain decomposition**



#### Communication



Setting up for distributed/multi-GPU computing
 Top level domain decomposition of the solution x
 One domain for each MPI thread / GPU



# We communicate the FP once per block Note that only part of the FP is required from others

#### Multi-GPU results (K80)





Ν	256	512	1024	2048
1 GPU	3.33	22.8	181	-
2 GPU	2.31	12.6	95.6	798
4 GPU	2.08	7.68	51.1	398
8 GPU	2.00	4.65	26.4	211

Walnut case - full reconstruction time (2 iters) without disk I/O in seconds.

Solution size: N<sup>3</sup> / Projections: 400xN<sup>2</sup>



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#### **GPU cluster results**







#### CPU cluster results (Xeon 2680)





Ν	256	512	1024	2048
20 cores	5.54	38.7	316	2380
40 cores	3.39	21.8	171	1260
80 cores	1.94	13.3	97,3	700
160 cores	1.36	9.55	60.1	413
320 cores	-	8.01	46.5	314

Walnut case - full reconstruction time (2 iters) without disk I/O in seconds.

Solution size: N<sup>3</sup> / Projections: 400xN<sup>2</sup>

### Summary



- Block methods are convenient for large-scale reconstructions where runtimes are an issue
- ASTRA allows you to do block methods with little effort and good performance for limited blocks
- Block methods are well suited for GPUs & CPUs
  - We have an implementation with one projection per block that performs close to SIRT runtimes
- Multi-GPU approach is also efficient by domain decomposition and communication of the FP
- GPU-cluster performance scales but operates at the limit of current communication bandwidths





#### Thank you for your attention!

#### **Block method**

Formulated in terms of matrices (as yesterday)

**Basic Block-Iteration algorithm** 

```
\begin{aligned} \mathbf{x}^{(0)} &= \text{initial vector} \\ \text{for } k &= 0, 1, 2, \dots \\ \mathbf{x}^{(k,0)} &= \mathbf{x}^{(k)} \\ \text{for } l &= 0, 1, 2, \dots, \text{blks-1} \\ \mathbf{x}^{(k,l+1)} &= P_{\mathcal{C}} \left( \mathbf{x}^{(k,l)} + \lambda_k \mathbf{D}^{-1} \mathbf{A}_l^T \mathbf{M}_l^{-1} (\mathbf{b}_l - \mathbf{A}_l \mathbf{x}^{(k,l)}) \right) \\ \text{end} \\ \mathbf{x}^{(k+1)} &= \mathbf{x}^{(k,\text{blks})} \\ \text{end} \end{aligned}
```