



GPU and  
Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

Conclusions

Journée GNR MOMAS / GDR Calcul

ENS ULM – PARIS

*Le projet BigDFT : Méthodes Ab initio,  
ondelettes et supercalculateurs massivement  
parallèles avec GPU*

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European Synchrotron Radiation Facility

5 May 2010

with M. Ospici, B. Videau, J.-F. Méhaut, T. Deutsch



- 1 Electronic structure calculations
  - Ab initio methods
  - BigDFT code
  - Main operations, parallelisation

- 2 Hybrid code
  - The S\_GPU library

- 3 Performances

- 4 Conclusions

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

Conclusions

# Ab initio calculations with DFT

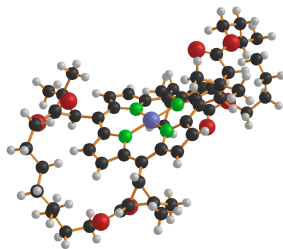
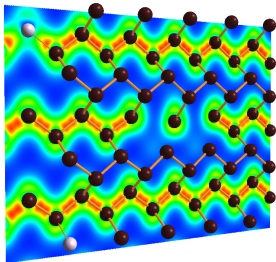
## Several advantages

- ✓ **Ab initio:** No adjustable parameters
- ✓ **DFT:** Quantum mechanical (fundamental) treatment

## Main limitations

- ✗ Approximated approach
- ✗ Requires high computer power, limited to few hundreds atoms in most cases

Wide range of applications: nanoscience, biology, materials



GPU and  
Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

Conclusions

# Performing a DFT calculation (KS formalism)

Find a set of orthonormal orbitals  $\Psi_i(\mathbf{r})$  that minimizes:

$$E = -\frac{1}{2} \sum_{i=1}^{N/2} \int \Psi_i^*(\mathbf{r}) \nabla^2 \Psi_i(\mathbf{r}) d\mathbf{r} + \frac{1}{2} \int \frac{\rho(\mathbf{r})\rho(\mathbf{r}')}{|\mathbf{r}-\mathbf{r}'|} d\mathbf{r}' d\mathbf{r} \\ + E_{xc}^{LDA}[\rho(\mathbf{r})] + \int V_{ext}(\mathbf{r})\rho(\mathbf{r})d\mathbf{r}$$

with  $\rho(\mathbf{r}) = \sum_i \Psi_i^*(\mathbf{r})\Psi_i(\mathbf{r})$

## (Kohn-Sham) DFT “Actors”

- A set of **wavefunctions**  $|\Psi_i\rangle$ , one for each electron
  - A computational approach on a **finite basis**
- ⇒ One **array** for each  $\Psi_i$
- ⇒ A set of **computational operations** on these arrays which depend on the basis set
- A (good) computer...



GPU and  
Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

Conclusions

# A basis for nanosciences: the BigDFT project



GPU and  
Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

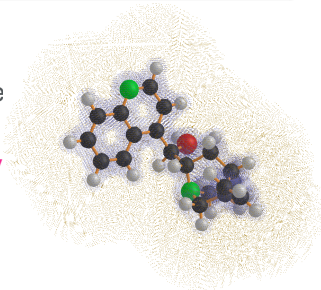
Conclusions

STREP European project: BigDFT(2005-2008)

Four partners, 15 contributors:

CEA-INAC Grenoble, U. Basel, U. Louvain-la-Neuve, U. Kiel

Aim: To develop an ab-initio DFT code based on **Daubechies Wavelets**, to be *integrated in ABINIT*, distributed **freely** (GNU-GPL license)



**L. Genovese**, A. Neelov, S. Goedecker, T. Deutsch, *et al.*,

“Daubechies wavelets as a basis set for density functional pseudopotential calculations”,

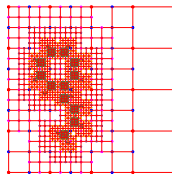
J. Chem. Phys. **129**, 014109 (2008)

# A DFT code based on Daubechies wavelets

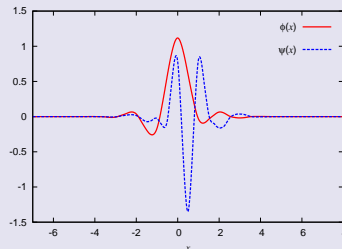
## Wavelets

A basis with optimal properties for expanding localised information

- Localised in real space
- Smooth (localised in Fourier space)
- Orthogonal basis
- Multi-resolution basis
- Adaptive
- Systematic



## Daubechies Wavelets



From early 80's

Applied in several domains

Interesting properties for DFT



GPU and  
Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

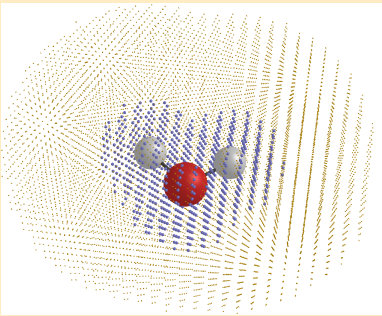
Performances

Conclusions

# Wavelet properties: adaptivity

## Adaptivity

Resolution can be refined following the grid point.



The grid is divided in **Low** (1 DoF) and **High** (8 DoF) resolution points. Points of different resolution belong to **the same** grid. Empty regions must not be “filled” with basis functions.

Localization property, real space description

Optimal for **big & inhomogeneous** systems, **highly flexible**



GPU and  
Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

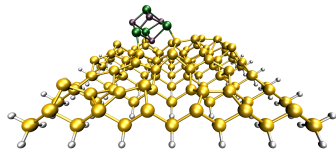
Performances

Conclusions

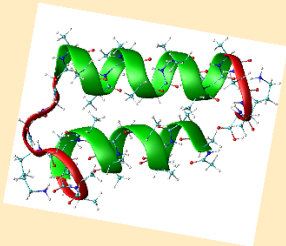
# Some ongoing applications

(Group of S. Goedecker, Basel University)

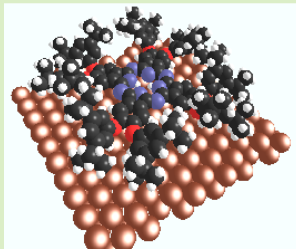
Study of favorable adsorption sites of NaCl clusters on Si tips during AFM experiments ( $\sim 250$  at.)



350 at.: DFT calculation to assess the accuracy of force fields



$\sim 300+500$  at. : Organic molecule @ Cu surface. HOMO-LUMO chg. dens. are compared to STM expts.



GPU and Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

Conclusions



# Operations performed

Different numerical operations performed:

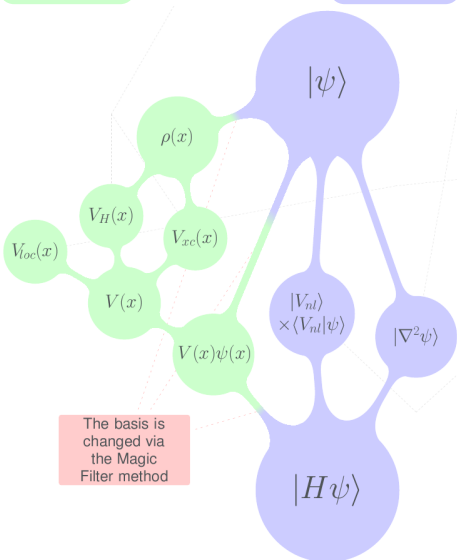
- For each wavefunction (Hamiltonian application)
- Between wavefunctions (Linear algebra)

## Comput. operations

- Convolutions with **short** filters
- BLAS routines
- FFT (Poisson Solver)

Interpolating

Daubechies



GPU and Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

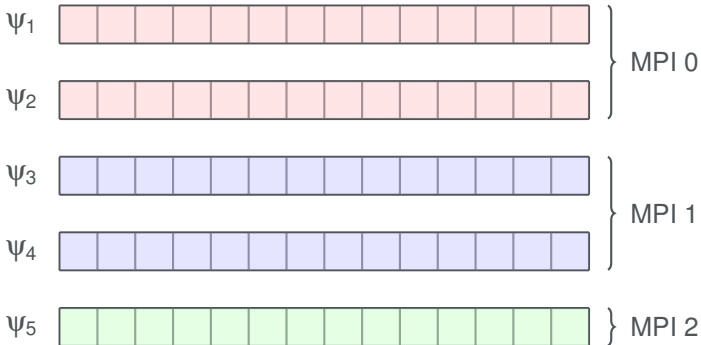
Performances

Conclusions

# Orbital distribution scheme

Used for the application of the hamiltonian

The hamiltonian (convolutions) is applied separately onto each wavefunction



GPU and Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

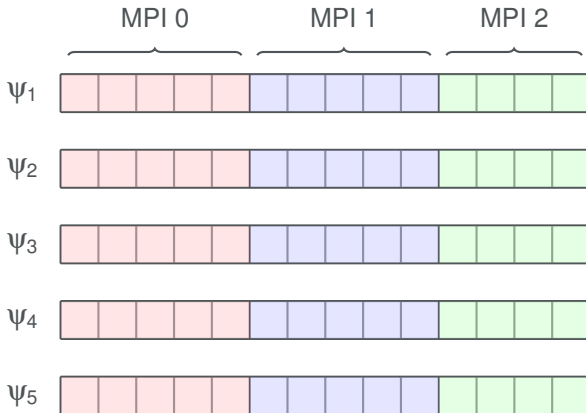
Performances

Conclusions

# Coefficient distribution scheme

Used for scalar product & orthonormalisation

BLAS routines (level 3) are called, then result is reduced



Communications are performed via **MPI\_ALLTOALLV**



GPU and  
Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

Conclusions

# High Performance Computing

## Localisation & Orthogonality → Data locality

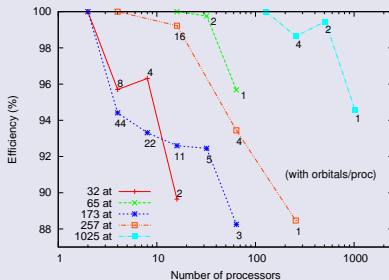
Principal code operations can be intensively optimised

## Optimal for application on supercomputers

Little communication, big packets of data

- No need of fast network
- Optimal speedup

Efficiency of the order of 90%, up to thousands of processors



## Data repartition optimal for material accelerators (GPU)

Graphic Processing Units can be used to speed up the computation



GPU and Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

Conclusions

# Separable convolutions



GPU and  
Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

Conclusions

We must calculate

$$\begin{aligned} F(l_1, l_2, l_3) &= \sum_{j_1, j_2, j_3=0}^L h_{j_1} h_{j_2} h_{j_3} G(l_1 - j_1, l_2 - j_2, l_3 - j_3) \\ &= \sum_{j_1=0}^L h_{j_1} \sum_{j_2=0}^L h_{j_2} \sum_{j_3=0}^L h_{j_3} G(l_1 - j_1, l_2 - j_2, l_3 - j_3) \end{aligned}$$

Application of three successive operations

- 1  $A_3(l_3, i_1, i_2) = \sum_j h_j G(i_1, i_2, l_3 - j) \quad \forall i_1, i_2;$
- 2  $A_2(l_2, l_3, i_1) = \sum_j h_j A_3(l_3, i_1, l_2 - j) \quad \forall l_3, i_1;$
- 3  $F(l_1, l_2, i_3) = \sum_j h_j A_2(l_2, l_3, l_1 - j) \quad \forall l_2, l_3.$

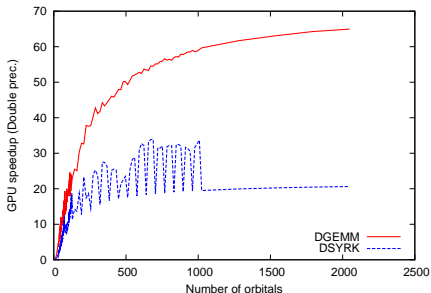
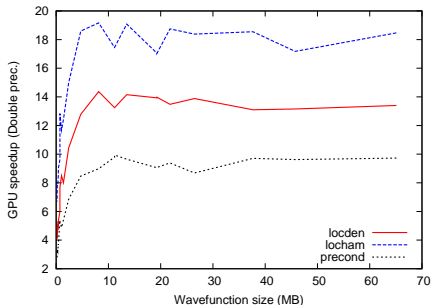
Main routine: Convolution + transposition

$$F(l, a) = \sum_j h_j G(a, l - j) \quad \forall a;$$

# GPU-ported operations in BigDFT (double precision)

## Convolutions (CUDA rewritten)

GPU speedups between 10 and 20 can be obtained for different sections



## Linear algebra (CUBLAS library)

The interfacing with CUBLAS is immediate, with considerable speedups



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Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

Conclusions

# Distribute the data on hybrid supercomputer



GPU and Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

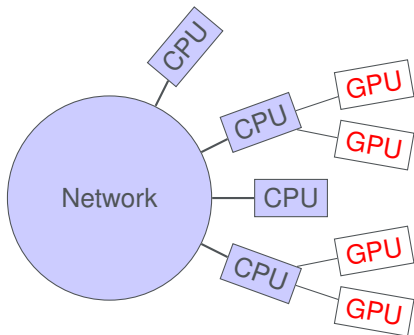
Hybrid code

S\_GPU

Performances

Conclusions

BigDFT should be executed on hybrid CPU-GPU architectures



Data transfer is still MPI-based

Only internode communication between GPU and CPU

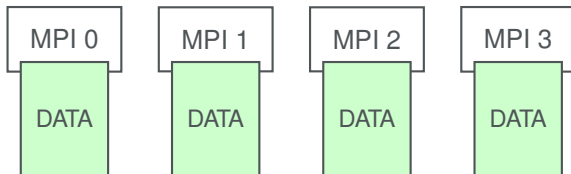
Data distribution should depend on the presence of GPUs on the nodes

# Data repartition on a node

## Non-hybrid case

GPU not used → homogeneous repartition

GPU



GPU and  
Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

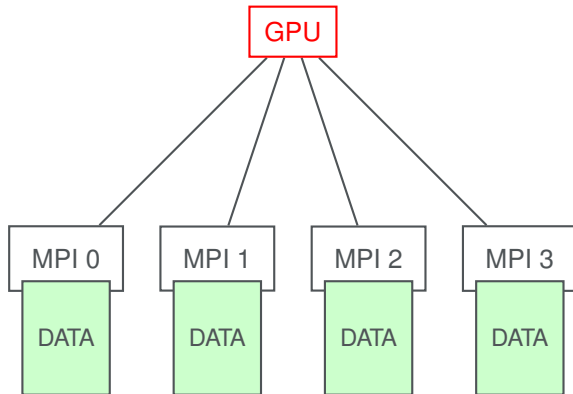
Conclusions



# Data repartition on a node

## “Naive” repartition

All the cores use the GPU at the same time



GPU and  
Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

Conclusions

# Data repartition on a node

## Inhomogeneous repartition

Only one node use the GPU with more data



GPU and Wavelets

BigDFT

Ab initio methods

BigDFT code

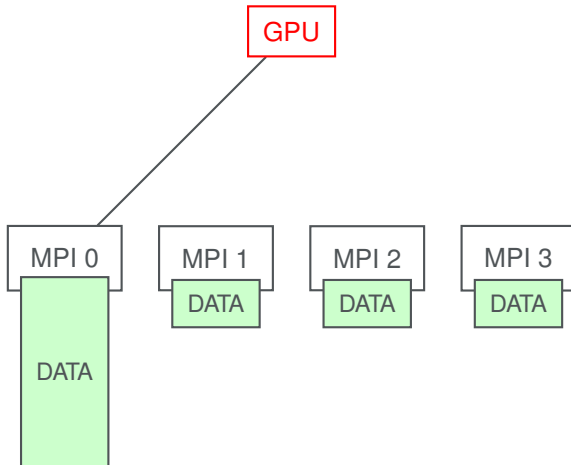
Parallelisation

Hybrid code

S\_GPU

Performances

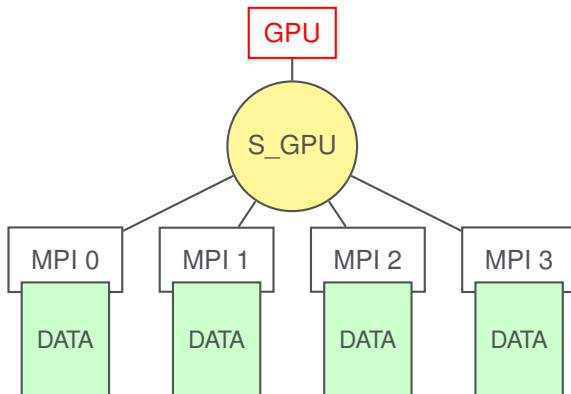
Conclusions



# Data repartition on a node

## The S\_GPU approach

S\_GPU library manages GPU resource within the node



GPU and  
Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

Conclusions



GPU and  
Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

Conclusions

## De-synchronisation of operations

Two semaphores are activated for each card on the node:

- Data transfer (CPU  $\rightarrow$  GPU and GPU  $\rightarrow$  CPU)
- Calculation on the GPU

Each operation (e.g. convolution of a wavefunction) is associated to a stream.

## Operation overlap

Calculation and data transfer of different stream may overlap  
Operation are scheduled on a first come - first served basis

## Several advantages

- The time for memory transfers is saved
- Heavy calculation can be passed to the card one - by - one, avoiding scheduling problems

# Example of a time chart



GPU and Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

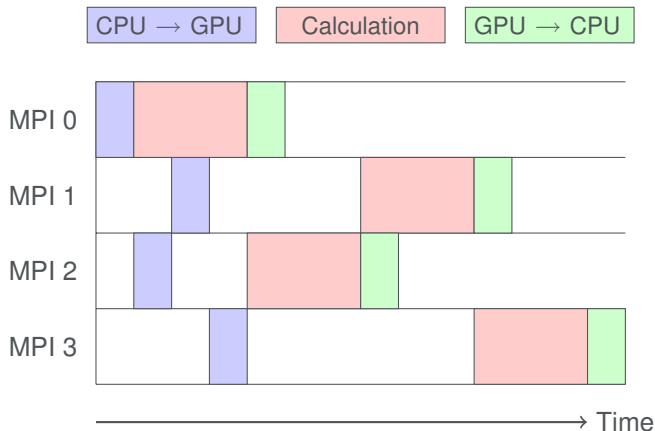
Hybrid code

S\_GPU

Performances

Conclusions

The GPU can be viewed as a **shared co-processor**



# Convenience of S\_GPU approach

## Different tests thanks to BigDFT flexibility

We have performed many tests, with different ratios GPU/CPU on the same node

## Speedup on the full code (exemples)

S\_GPU is the best compromise speedup/easiness

Examples:

CPU -GPU	8 - 1	8 - 2	4-2	2-2
S_GPU	1.96	3.69	3.73	5.09
Inhomogeneous (best)	2.08	2.64	2.32	2.40

## Full code tested on Multi-GPU platforms

- CINES -Iblis  
48 GPU, Prototype calculations
- CCRT - Titane  
Up to 196 GPU (Grand challenge 2009)



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BigDFT  
Ab initio methods  
BigDFT code  
Parallelisation

Hybrid code  
S\_GPU

Performances

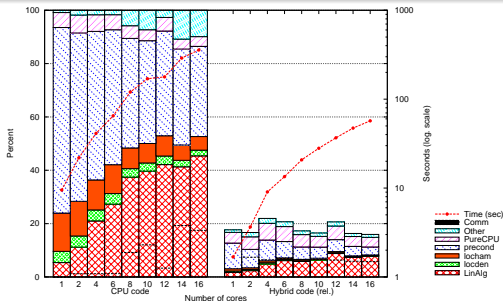
Conclusions

# BigDFT code on Hybrid architectures

BigDFT code can run on hybrid CPU/GPU supercomputers  
In multi-GPU environments, **double precision** calculations

## No Hot-spot operations

Different code sections can be ported on GPU  
up to 20x speedup for some operations,  
7x for the full parallel code (under improvements)



# BigDFT code on Hybrid architectures



GPU and  
Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

Conclusions

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## LG: Prix Bull-Fourier 2009 (Bull-GENCI)



Reference Paper: LG *et al.*, J. Chem. Phys. **131**, 034103 (2009)  
Grand Challenge 2009 on CEA Hybrid Cluster "Titane" (192  
GPU, 768 Intel Xeon Nehalem cores)



# Conclusions



GPU and  
Wavelets

BigDFT

Ab initio methods

BigDFT code

Parallelisation

Hybrid code

S\_GPU

Performances

Conclusions

## BigDFT code: a modern approach for nanosciences

- ✓ Flexible, reliable formalism (wavelet properties)
- ✓ Conceived for massive parallel architecture
- ✓ Open a path toward the diffusion of Hybrid architectures

## Present developments

- OpenCL BigDFT code (no performance loss)
- Use new drivers (concurrent kernel execution)
- New conception of S\_GPU

## BigDFT 1.4 beta – GNU-GPL license

Lots of applications & developments with BigDFT team:  
D. Caliste, T. Deutsch ([L\\_Sim - CEA INAC Grenoble](#))  
S. Goedecker ([U. Basel](#))  
M. Ospici, J-F. Méhaut ([LIG INRIA UJF Bull Grenoble](#))