

# Performance Optimization of CPMD

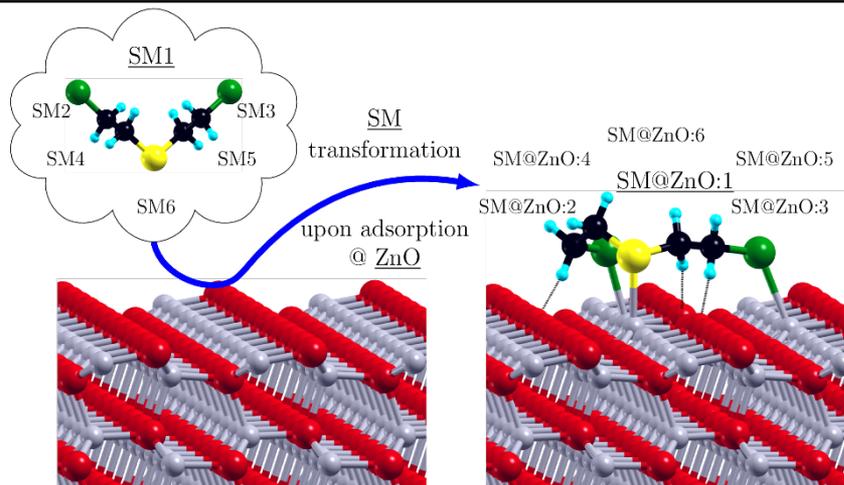
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Friedrich-Alexander-Universität Erlangen-Nürnberg

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at Erlangen Regional Computing Center (RRZE)  
Friedrich-Alexander-Universität Erlangen-Nürnberg

[3] Leibniz Supercomputing Centre (LRZ), Garching

# Static Simulations: Adsorption on Surfaces



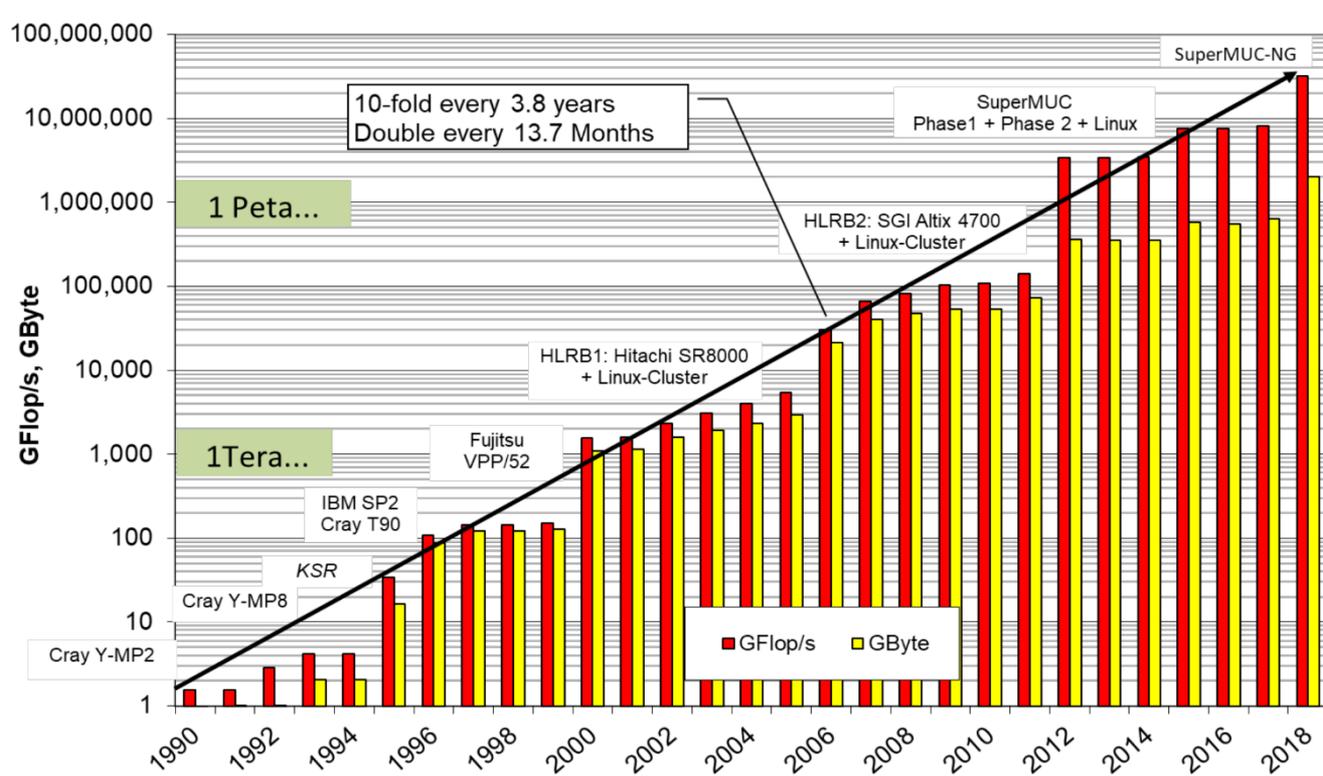
- 6 most stable gas phase conformers probed
- 279 independent geometry optimizations
- 150-250 geometry updates per geometry optimization
- 20 wave-function updates per geometry update  
-> 1E7 wave-function & 1E6 force updates

# Dynamic Simulations: Molecules in Solution

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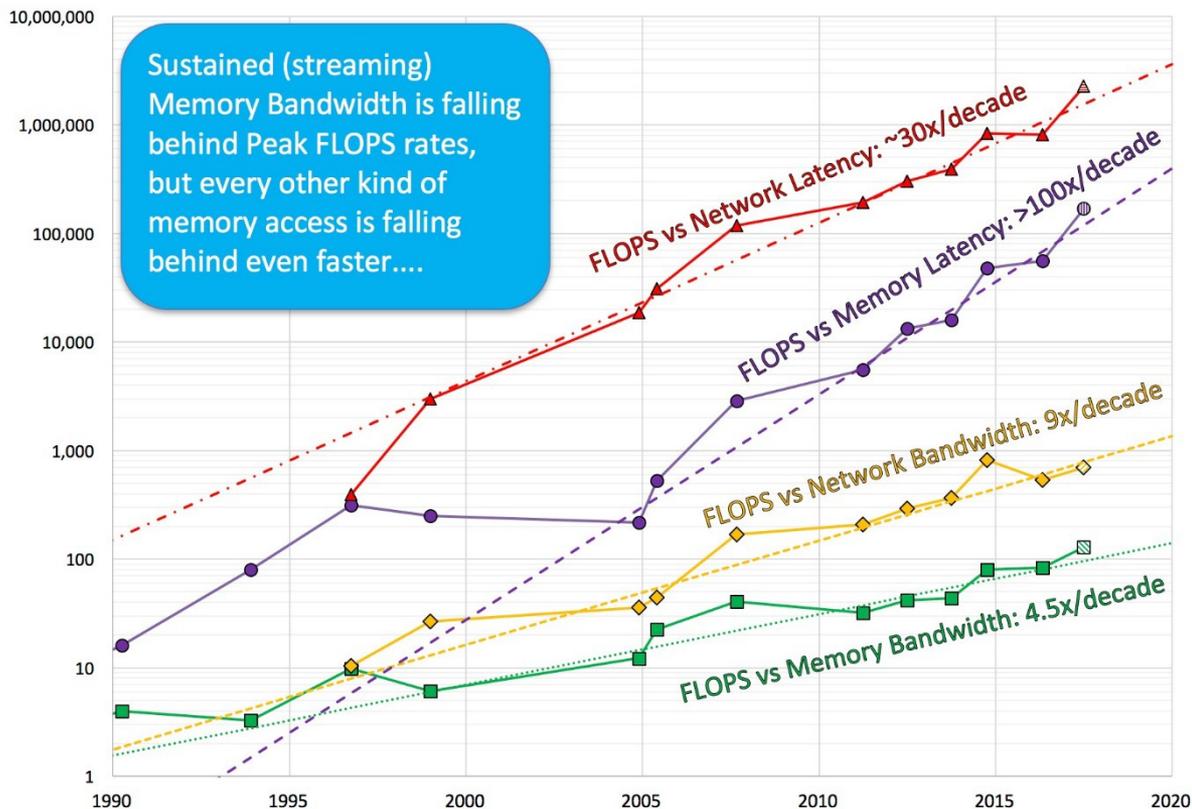
- 50 ps equilibration
- 250 ps simulated time
- 0.145 fs time step -> 20 M force **and** wave-function updates
- Multiple trajectories mandatory!
- System size may increase (cubic scaling DFT!)
- Parallelization of time not possible
- Vastly different computational requirements  
-> Extremely efficient code needed

# Moore's Law @ LRZ



Linpack performance!  
~ DGEMM performance

# Moore's Law @ LRZ



McCalpin, SC16: <http://sc16.supercomputing.org/wp-content/uploads/2016/10/McCalpin.jpg>

# Car-Parrinello Molecular Dynamics (CPMD) Code

Schrödinger equation in the framework of Density Functional Theory

**basis set: plane waves + pseudopotentials**

Pros:

- No Pulay forces
- No basis set superposition errors
- Single parameter to tune basis set size
- Periodic
- **FFTs for G/R space transformations**

Cons:

- Isolated systems
- Expensive vacuum
- Core electrons

# Pseudopotential Approach

## Normconserving NC-PP Pseudopotentials

- Many plane waves
- Typical 3D-FFT grid size: 200 ... 400
- **Thoroughly optimized by IBM Research (Rüschlikon)**
- **Dominated by 3D-FFTs**

## Ultrasoft US-PP Vanderbilt pseudopotentials

- Less plane waves
- Typical 3D-FFT grid size: 100 ... 200
- Approx. 10x less work in 3D-FFT!
- Overhead:  
 $\langle \Phi | \Phi \rangle \rightarrow \langle \Phi | S | \Phi \rangle!$
- Transformation of overhead into DGEMMs?

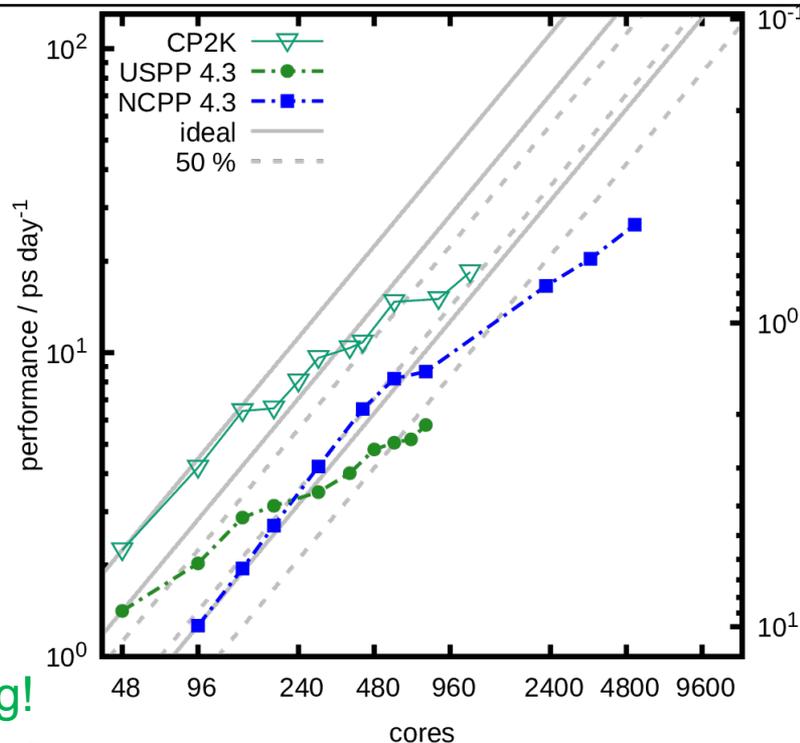
# CPMD Strong Scaling: Starting Point

## NC-PP:

- $N_\beta = 256$
- FFT:  $216^3$  (80 Ry)

## US-PP:

- $N_\beta = 2560$
- FFT:  $120^3$  (25 Ry)



## NC-PP

- superlinear scaling!
- Best time to solution
- CP2K much more efficient!

## SuperMUC-NG:

- Intel® Skylake Xeon Platinum 8174 (48 cores / node)
- Fully nonblocking fat tree Intel® OmniPath

## US-PP

- High performance at low core counts
- No OpenMP/MPI hybrid parallelization

# CPMD Internal Instrumentation

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SUBROUTINE	CALLS	SELF TIME		TOTAL TIME	
		AVERAGE	MAXIMUM	AVERAGE	MAXIMUM
cpmd	1	0.17	0.18	108.66	108.67
mdpt	1	0.02	0.02	107.96	107.96
mdmain	1	0.38	0.70	107.94	107.94
forcedr	51	0.01	0.01	98.24	98.24
noforce	51	1.87	2.28	98.23	98.24
rnism	102	0.00	0.00	37.62	37.74
rnism2	51	22.00	22.91	29.63	29.74
rscpot	51	0.01	0.01	21.30	21.32
vpsi	51	3.39	3.43	14.81	14.83
rhoofr	51	1.97	2.04	14.53	14.53
invfftn	51	11.59	11.64	11.59	11.64
fwfftn	51	11.42	11.44	11.42	11.44
nlforce	51	9.05	9.28	9.05	9.28
rnism2_b	306	7.63	8.41	7.63	8.41
rnism1	51	5.95	6.21	7.99	8.01
vofrho	51	0.00	0.00	5.32	5.33
rotate	85	4.06	4.21	4.06	4.21
ovlap	103	3.95	4.08	3.95	4.08

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COMMUNICATION TASK	AVERAGE MESSAGE LENGTH	NO. CALLS
SEND/RECEIVE	54689. BYTES	19476.
BROADCAST	814689. BYTES	595.
GLOBAL SUMMATION	2981743. BYTES	2049.
ALL TO ALL COMM	273449. BYTES	52785.
ALLGATHERV	2615031. BYTES	409.

	PERFORMANCE	TOTAL TIME
SEND/RECEIVE	3833.006 MB/S	0.278 SEC
BROADCAST	2310.022 MB/S	0.210 SEC
GLOBAL SUMMATION	2000.989 MB/S	13.196 SEC
GLOBAL MULTIPLICATION	0.000 MB/S	0.001 SEC
ALL TO ALL COMM	1333.118 MB/S	10.827 SEC
ALLGATHERV	480.392 MB/S	2.226 SEC
SYNCHRONISATION		0.510 SEC

- Timings excluding/including subroutines
- Communication heavy on global summation and all-to-all communication

# How Can We Improve the US-PP Code Path?

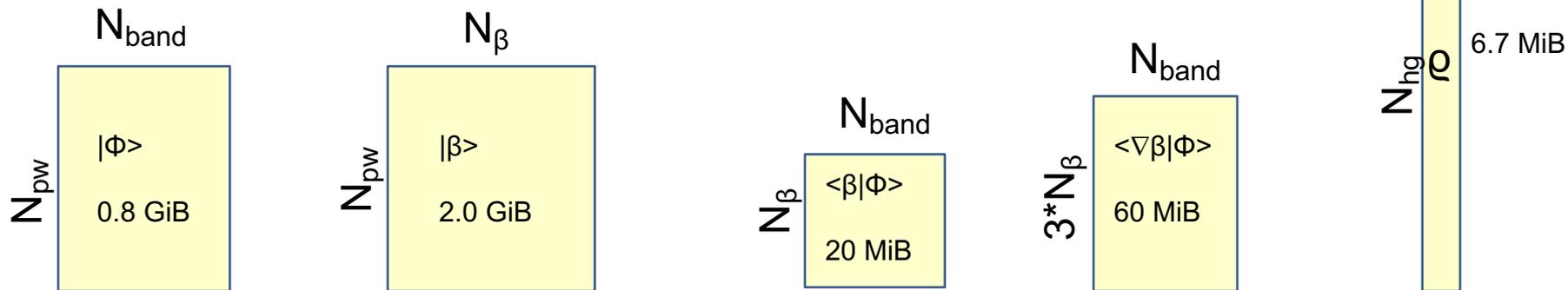
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- Understand data structures
- Understand node level performance
- Understand MPI performance

# Data Structures

256 H<sub>2</sub>O molecules, 25 Ry wave-function cutoff, 100 Ry charge-density cutoff, 19.73 Å<sup>3</sup>

- 120<sup>3</sup> FFT grid
- $N_{pw} = 54564$  plane wave coefficients for wave-functions in G-space
- $N_{hg} = 437792$  plane wave coefficients for charge-density in G-space
- $N_{band} = 1024$  bands
- $N_{\beta} = 2560$   $\beta$ -projectors

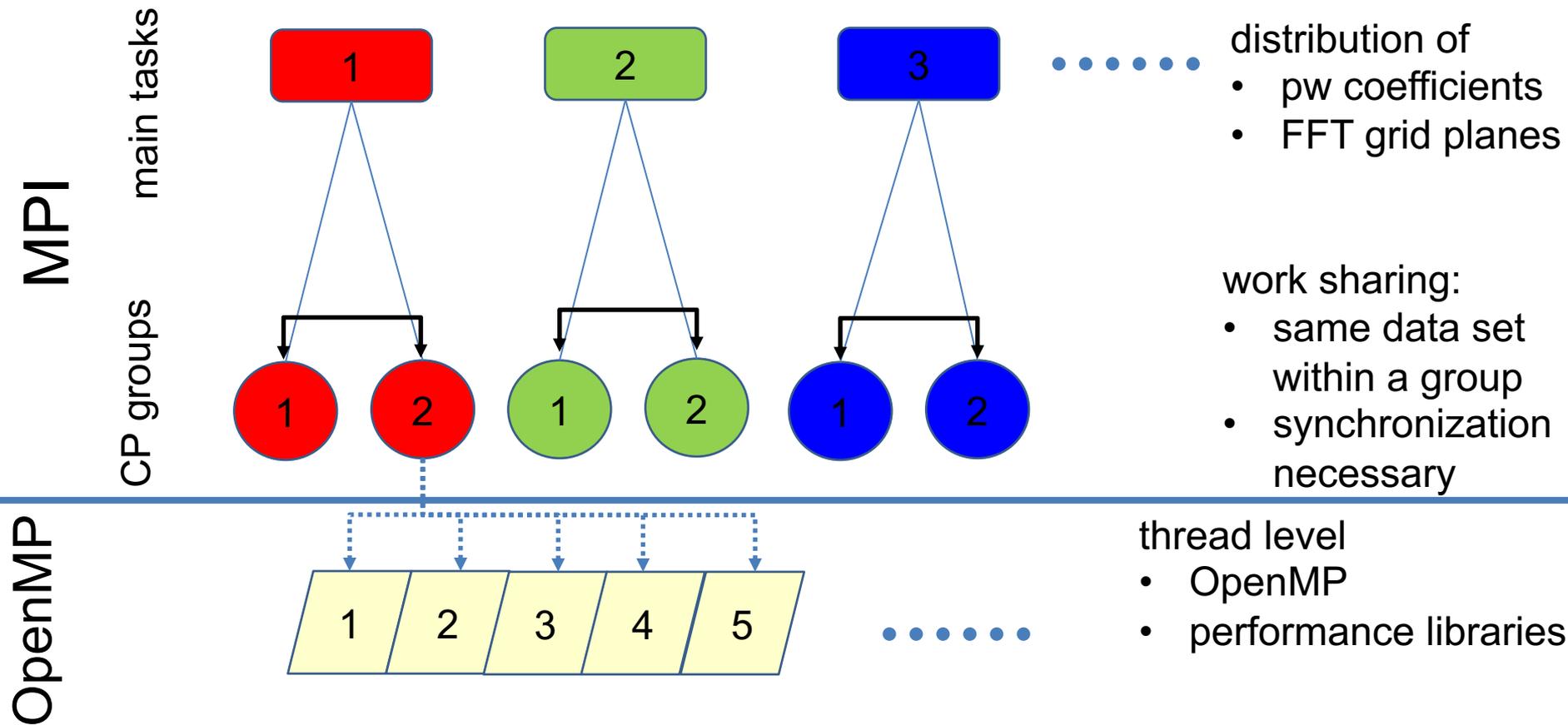


# Parallelization Strategies in CPMD

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1. Distribution of  $N_{pw}$  plain waves (basis functions) across main MPI tasks
2. Second level MPI parallelization with so called `cp_groups`:
  - 2 (or few) communicators with replicated data
  - parallelization over  $N_{band}$  electronic states
  - parallelization over  $N_{\beta}$  projectors
  - implemented only for selected routines along the main code path
3. Thread parallelization with OpenMP and threaded performance libraries
  - efficient only within NUMA domains.
  - implemented only for selected routines along the main code path

# Parallelization Layers of CPMD



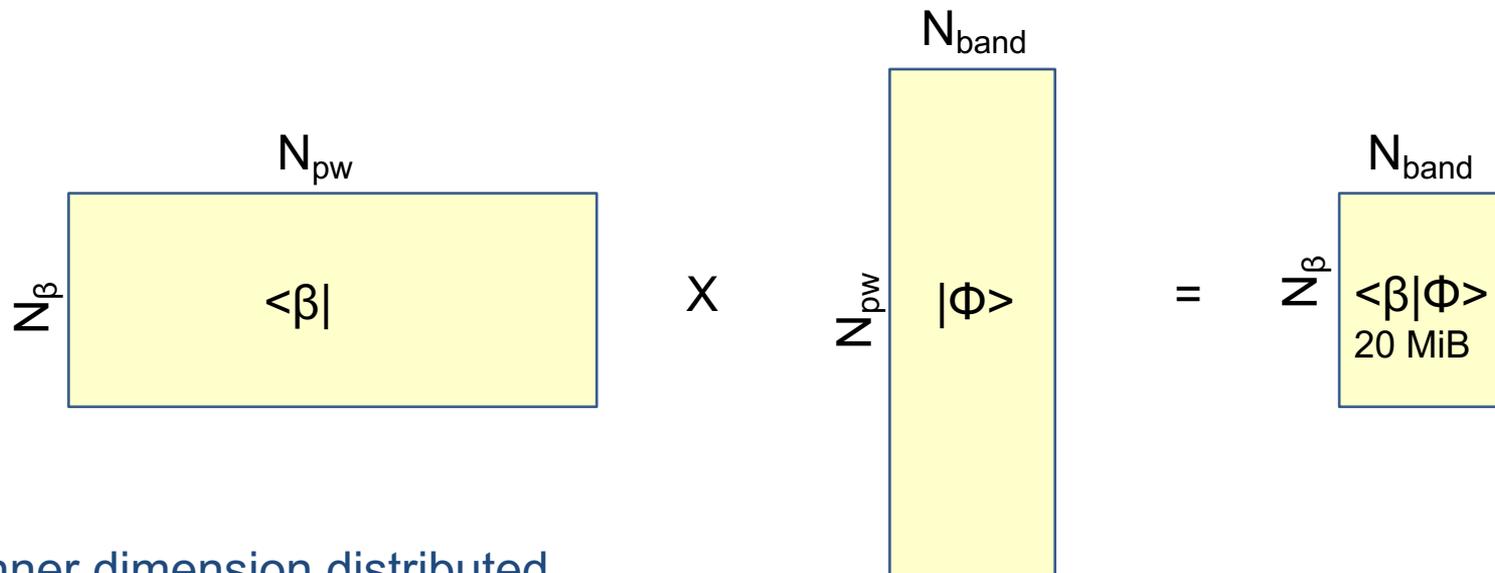
# Distribution of Plane Wave Coefficients

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1. **Distributed Matrix Matrix Multiplication ( $\langle \beta | \Phi \rangle$  and  $\langle \nabla \beta | \Phi \rangle$ )**
2. Distributed 3D-FFT transformation ( $|\Phi\rangle_G \rightarrow |\Phi\rangle_R$ )

# Distributed Matrix Matrix Multiplication

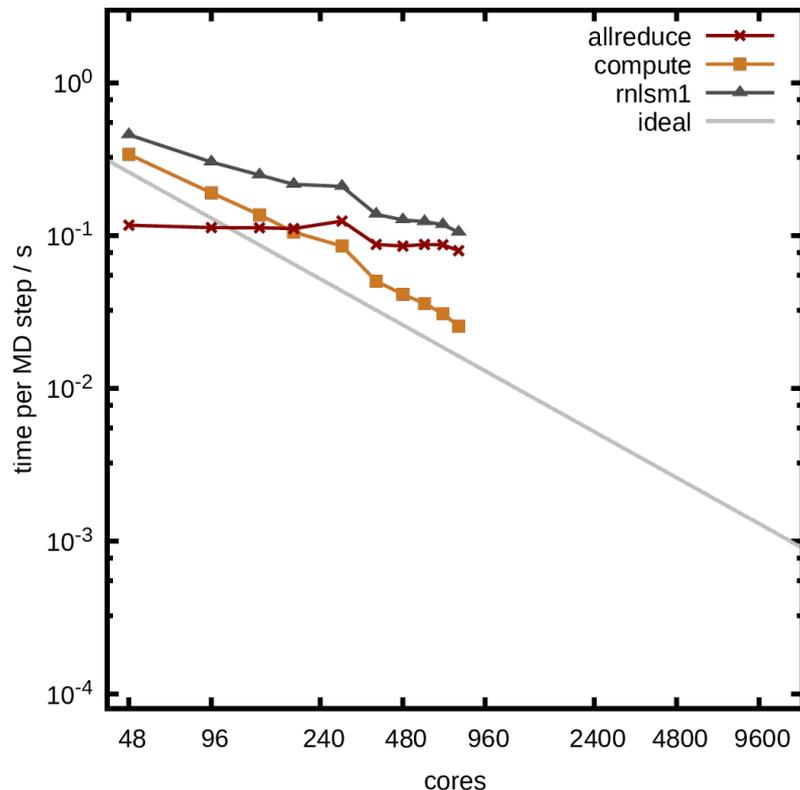
Calculation of  $\langle\beta|\Phi\rangle$



- Inner dimension distributed
- $\langle\beta|\Phi\rangle$  is replicated at each MPI task!
- Local DGEMMs + MPI\_allreduce of 20 MiB

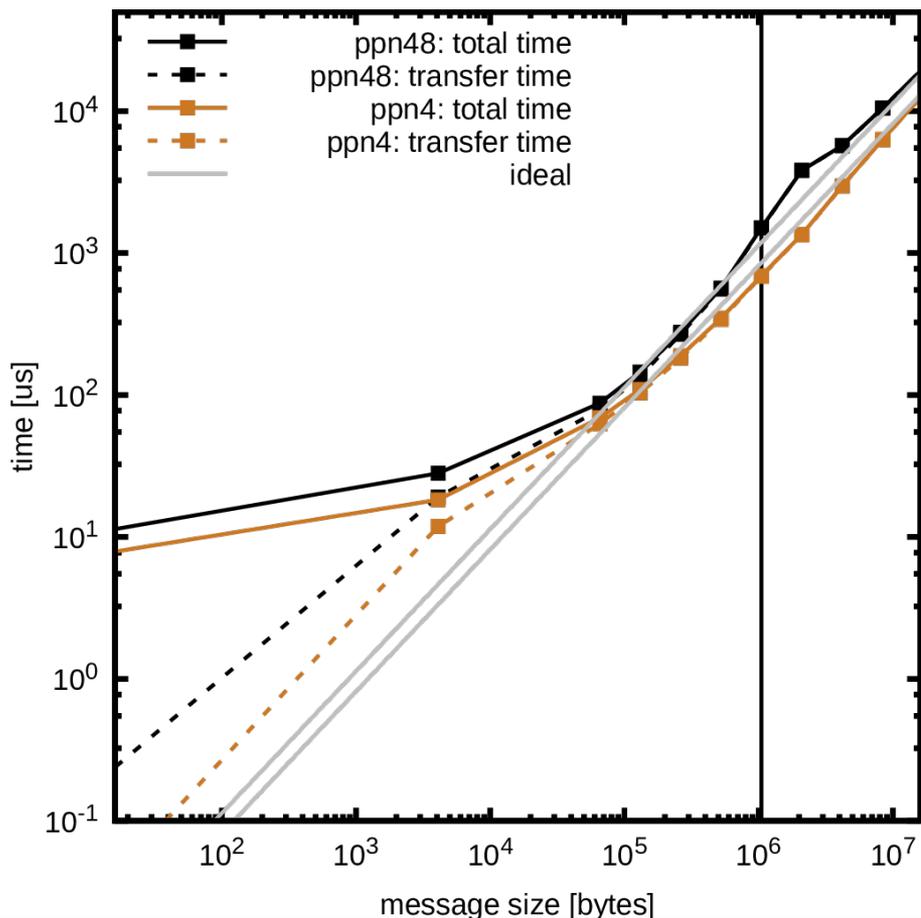
# Calculation of $\langle \beta | \Phi \rangle$

One DGEMM + MPI\_allreduce call for each atomic species and each  $\beta$ -projector



- Number of MPI tasks / OpenMP threads according to overall best performance!
- At 16 nodes (ppn8, 768 cores):
  - MPI comm: 0.080 s/MD
  - Compute: 0.026 s/MD
- MPI\_allreduce could benefit from larger message size
- DGEMMs should be kept as big as possible (and as quadratic as possible)

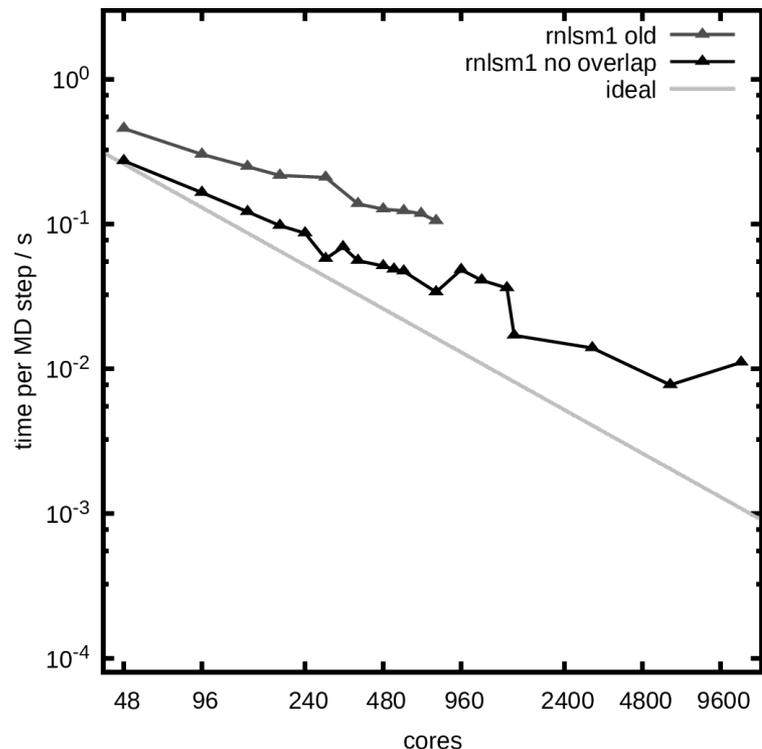
# MPI Allreduce Performance @ 16 Nodes



- Almost no benefit from using fewer MPI ranks
- Allreduce size should be >512 KiB

# Distributed Matrix Matrix Multiplication ( $\langle\beta|\Phi\rangle$ )

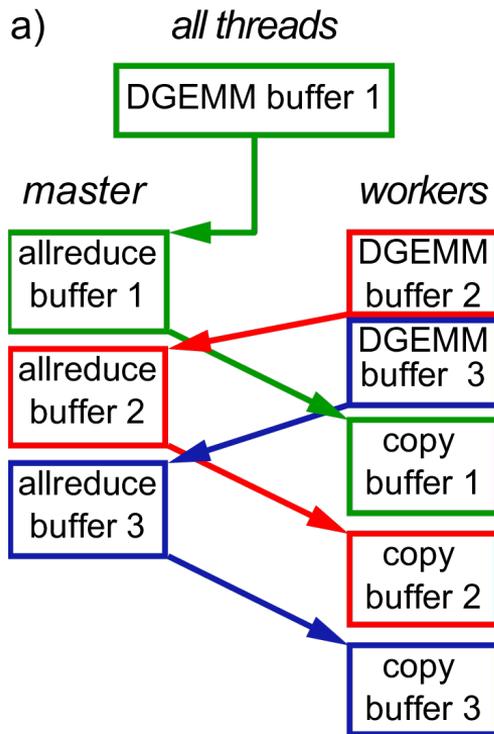
Idea 1: Use a single DGEMM + MPI\_allreduce (all species all projectors)



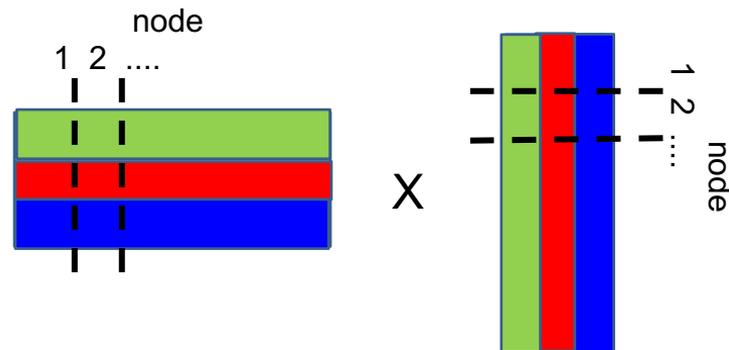
- Number of MPI tasks / OpenMP threads according to overall best performance!
- Total time at 16 nodes: 0.034 s/MD step
- Distribute  $\beta$ -projectors across cp\_groups
- cp\_groups active at  $\geq 1536$  cores
- cp\_groups overhead not shown!

# Distributed Matrix Matrix Multiplication ( $\langle \beta | \Phi \rangle$ )

Idea 2: Overlap of communication and computation



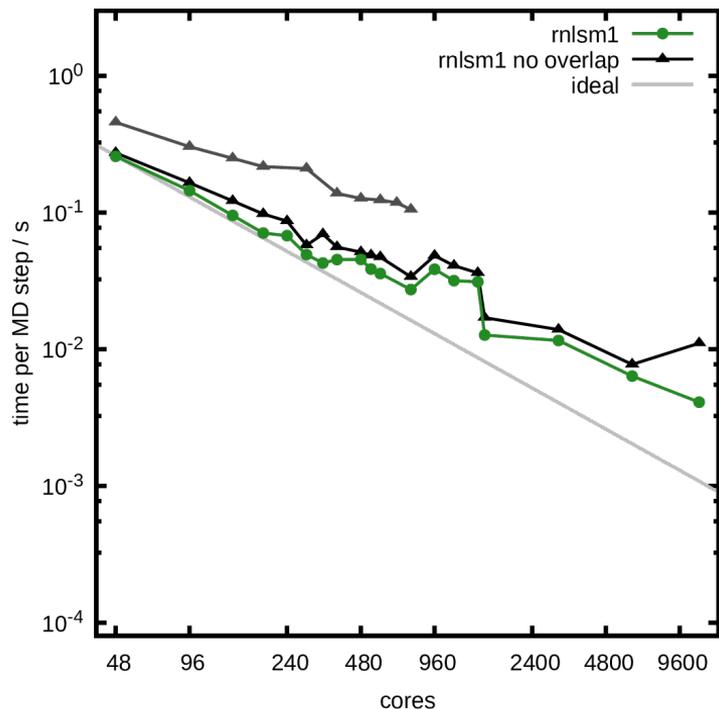
- Split DGEMM into few smaller parts
- OpenMP master thread used for communication
- OpenMP threads 2:n for remaining DGEMMs (nested OpenMP parallelism!)



# Distributed Matrix Matrix Multiplication ( $\langle\beta|\Phi\rangle$ )

Idea 1: Use a single DGEMM + MPI\_allreduce (all species all projectors)

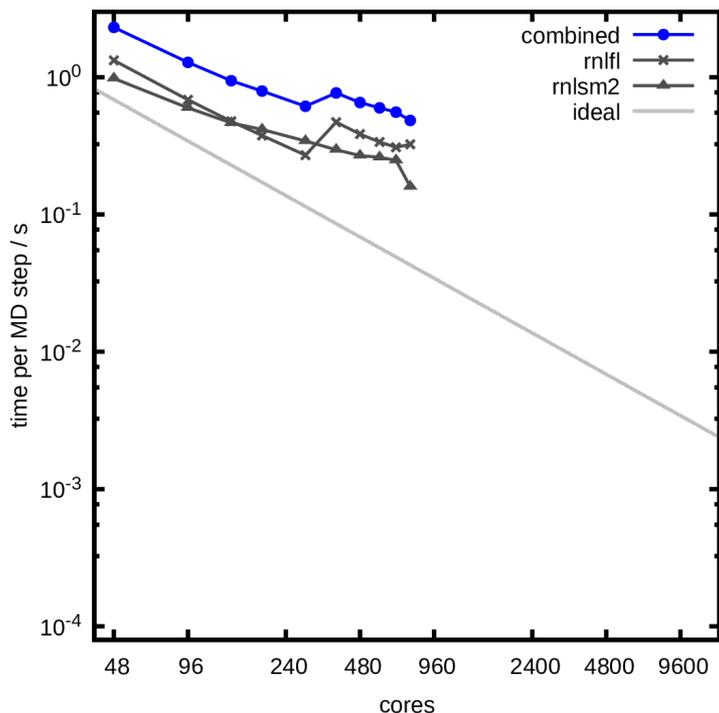
Idea 2: Overlap of communication and computation



- Total time at 16 nodes:  
0.034 s/MD step
- Total time (overlap) at 16 nodes:  
0.027 s/MD step
- Speedup at 16 nodes  
(old vs new + overlap):  
3.9

# Distributed Matrix Matrix Multiplication ( $\langle \nabla\beta|\Phi\rangle$ )

Idea 1: apply same optimizations as for  $\langle\beta|\Phi\rangle$   
+ optimization of rnlfl (hidden DGEMM ( $\langle\beta|\Phi\rangle \times \langle\Phi|H|\Phi\rangle$ ))

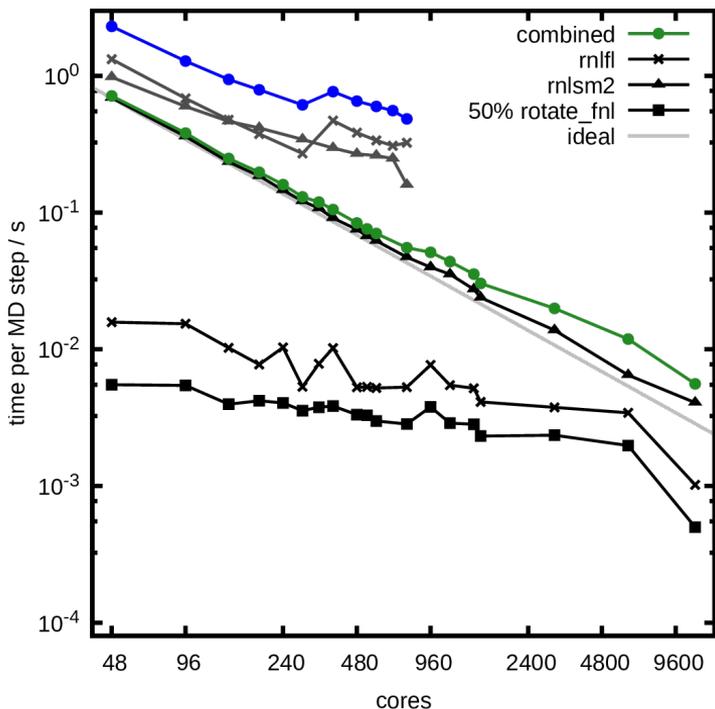


- No OpenMP inside rnlfl
- Only rnlfl needs  $\langle\nabla\beta|\Phi\rangle$  (ionic forces)

# Distributed Matrix Matrix Multiplication ( $\langle \nabla \beta | \Phi \rangle$ )

Idea 1: apply same optimizations as for  $\langle \beta | \Phi \rangle$   
+ optimization of rnlfl (hidden DGEMM ( $\langle \beta | \Phi \rangle \times \langle \Phi | H | \Phi \rangle$ ))

Idea 2: Discard MPI\_Allreduce



- Optimized rnlfl discards parallelization
- rottr\_fnl (DGEMM) parallelized at node level only
- Discard MPI\_Allreduce
- Almost ideal scaling

# Distribution of Plane Wave Coefficients

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1. Distributed Matrix Matrix Multiplication ( $\langle \beta | \Phi \rangle$  and  $\langle \nabla \beta | \Phi \rangle$ )
2. **Distributed 3D-FFT transformation ( $|\Phi\rangle_G \rightarrow |\Phi\rangle_R$ )**

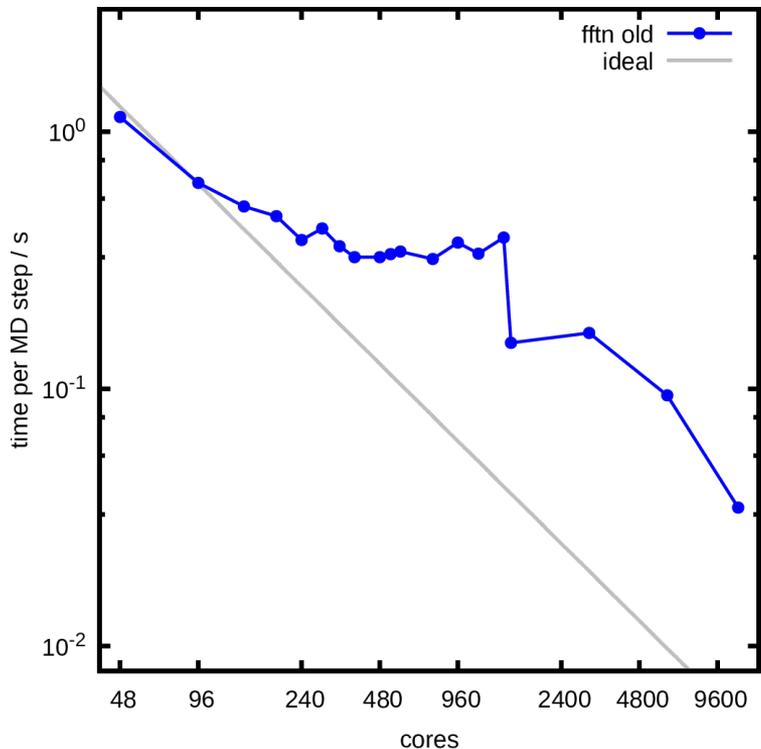
# Distributed 3D-FFT - Parallelization

- 3D-FFT for each of the  $N_{\text{band}}$  electronic states ( $\sim 120^3$  grid each)  
 $N_{\text{pw}}$  plane wave coefficients distributed over the MPI task
- Distribute planes in real space  
scaling limited to number of planes, here 120 MPI tasks  
48 cores per node @ LRZ, 128 cores per node @ HLRS
- Add more resources to a single MPI task for the actual 1D-FFT computations:  
hybrid parallelization ( MPI + X, X = OpenMP, accelerators, ...)
- cp\_group parallelization:  
distribute electronic states among cp\_groups  
data replication + synchronization

# Distributed 3D-FFT MPI + OpenMP

Already implemented in CPMD

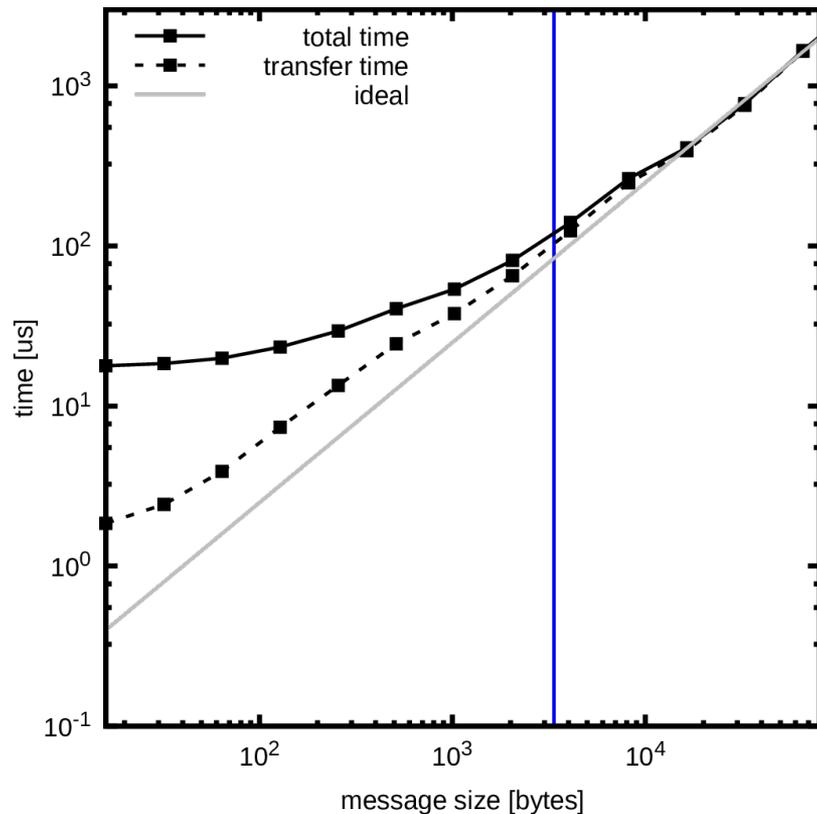
Performance of new US-PP implementation with old 3D-FFT routines



- Number of MPI tasks / OpenMP threads according to overall best performance!
- Scaling of FFT in hybrid parallelization: 240 cores
- Large performance benefit of using cp\_group parallelization at 1536 cores! (cp\_group overhead not shown!)

# MPI Alltoall Performance

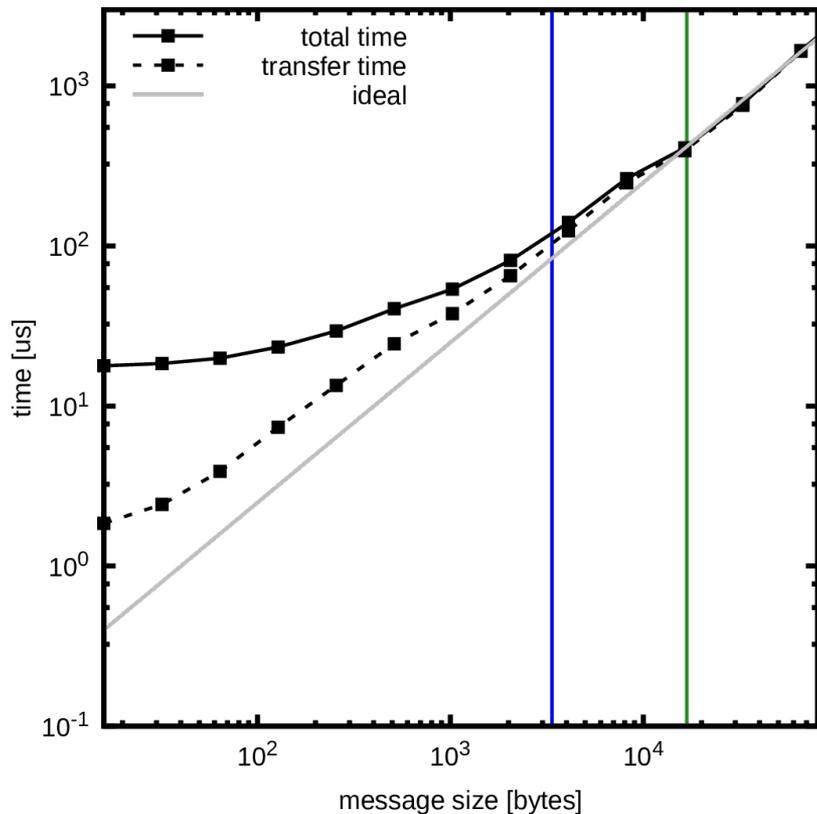
5 nodes, 8 MPI tasks per node



- All to all latency bound!
- FFT All-to-all message size 3360 bytes ( 3 planes x 70 rays )
- Message size will decrease with increasing MPI tasks

# MPI Alltoall Performance

## Idea 1: Combine A2A communication



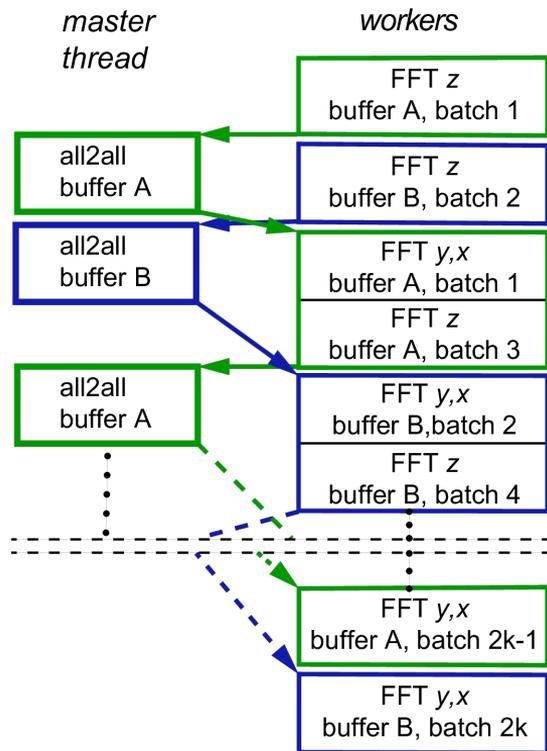
- Combine several all-to-all calls  
→ pack/unpack state information

States / A2A	1	2	3	4
tot. time [ms]	514	457	428	418
	5	6	7	8
	404	423	417	423

- total time decreases up to sweet spot (5 states)
- larger effect for higher node count expected

# MPI Alltoall Performance

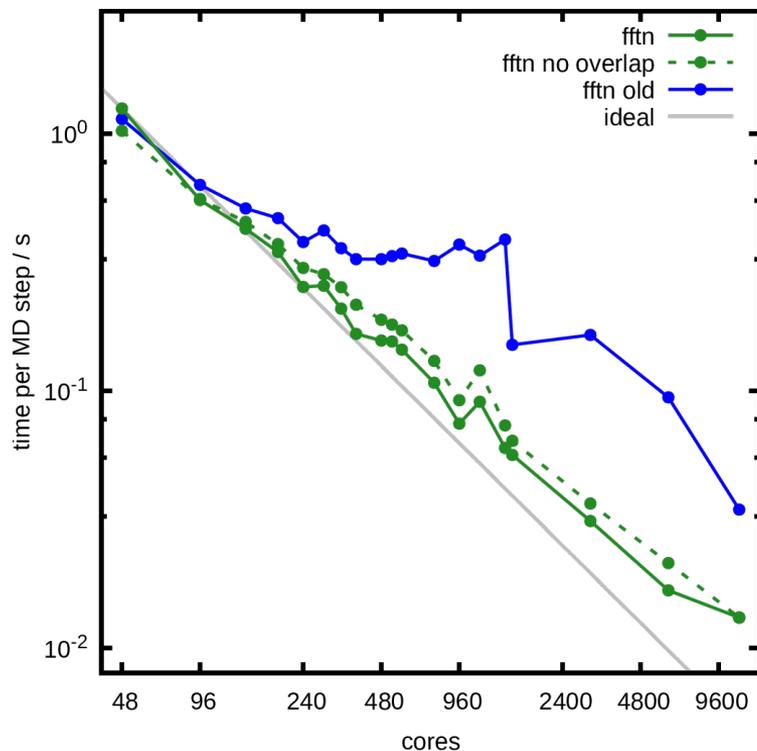
Idea 2: Work on two batches to hide communication



States A2A	Msg size [bytes]	Time [ms]	Time [ms] (overlap)
1	3360	514	504
2	6720	457	379
3	10080	428	358
4	13440	418	352
5	16800	404	371
6	20160	423	376
7	23520	417	403
8	26880	423	399
9	30240	439	407
10	33600	462	410

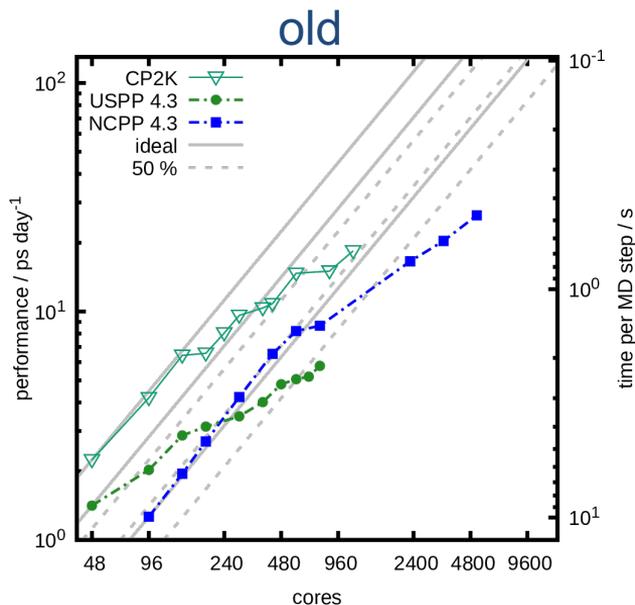
# New Batched 3D-FFT

Performance of new US-PP implementation with new 3D-FFT routines



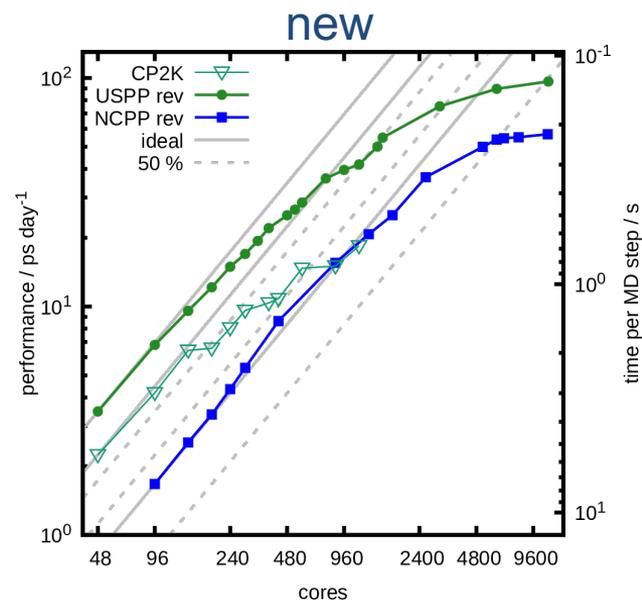
- Number of MPI tasks / OpenMP threads according to overall best performance!
- Scaling of FFT in hybrid parallelization: > 4800 cores
- No performance benefit of using `cp_group` parallelization at 1536!

# CPMD US-PP > 15,000 Codes Lines Changed



NC-PP

- 1.2x-1.3x speedup
- 50ps per day
- Outperforms CP2K

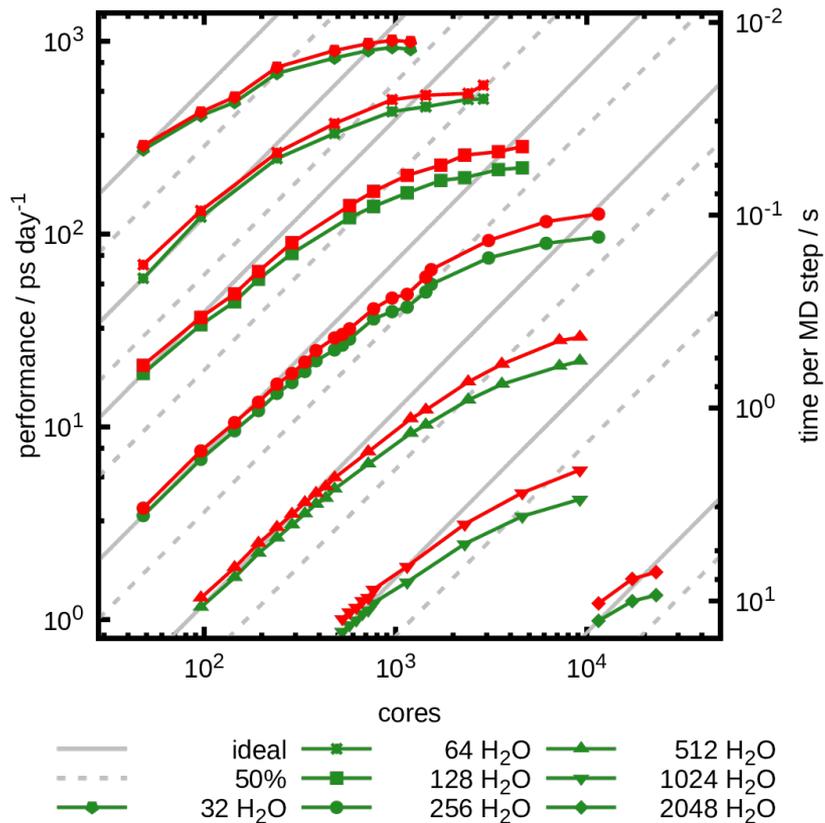


US-PP

- >2.0x speedup
- 70ps per day
- Best time to solution!
- Most efficient

# Strong Scaling Benchmark

32 – 2048 H<sub>2</sub>O Molecules @ SuperMUC-NG



- Optimized for 1400 – 3000 electrons
- Excellent performance also for tiny systems! More than **950 ps/day** -> QM/MM simulations
- If you really want to: affordable DFT calculation with 2048 H<sub>2</sub>O molecules, 16384 electrons! Code not even optimized

# Take Home Messages: Node Level Optimization

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- Check if you can map inner loops to BLAS calls.  
→ let the performance library do the work for you
- Thread parallelization is included
- Overhead to locally rearrange data (e.g. matrix buildup) is often acceptable (for BLAS level 2 or 3)
- Check, if you can combine smaller matrices to a larger one.  
→ better vectorization and less overhead
- Large BLAS operations ready for offloading
- Variants: Check batched BLAS (MKL), libxsmm (small matrices)

# Take Home Messages: Communication

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- Check ,if you are running into latency bound regimes on scale out
- Combine communication calls to stay in bandwidth bound regime
- Check, if communication can be avoided
  - Is the information really necessary (in all cases)?
  - Is MPI\_allreduce needed or is MPI\_reduce sufficient?
  - Is there a faster node local algorithm?
- Overlapping communication and computation can give you the last bleeding edge. (max. gain is a factor of 2)

## Read more ...

- Tobias Klöffel, Gerald Mathias, Bernd Meyer, Integrating state of the art compute, communication, and autotuning strategies to multiply the performance of ab initio molecular dynamics on massively parallel multi-core supercomputers, Computer Physics Communications, Volume 260, 2021, 107745, <https://doi.org/10.1016/j.cpc.2020.107745>.  
(<https://www.sciencedirect.com/science/article/pii/S0010465520303684>)
- Integrating State of the Art Compute, Communication, and Autotuning Strategies to Multiply the Performance of the Application Programm CPMD for Ab Initio Molecular Dynamics Simulations  
<https://arxiv.org/abs/2003.08477>

# Acknowledgements

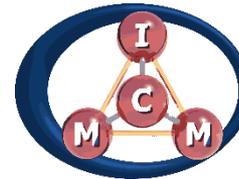
Tobias Klöffel



Bernd Meyer



Georg Hager

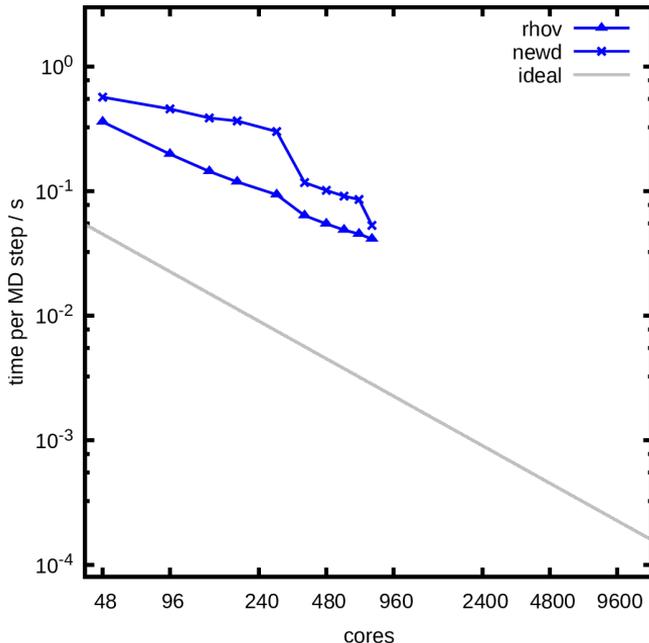


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# Backmatter

# Calculation of Augmentation Charges & New D

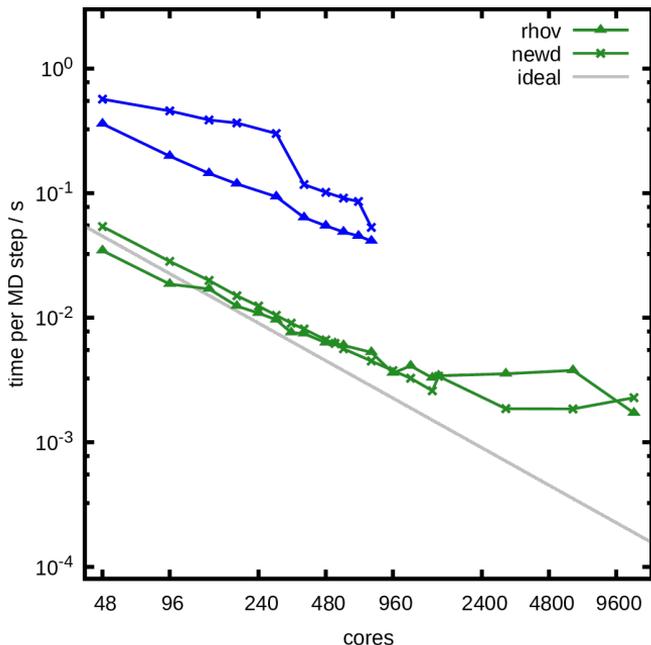
- Rhov: calculation of augmentation charges  
Newd: calculation of D, Q and ionic forces
- Recalculation of Q-function at every call in both routines
- Calculation of becsun in both routines, differently implemented, not parallelized



- Rhov  
DGEMV for each  $\beta$ -projector combination for each atomic species (37)
- Newd  
DGEMM for each atomic species (2)  
separate DGEMM for ionic forces (37)  
MPI summation of sparse array deeq

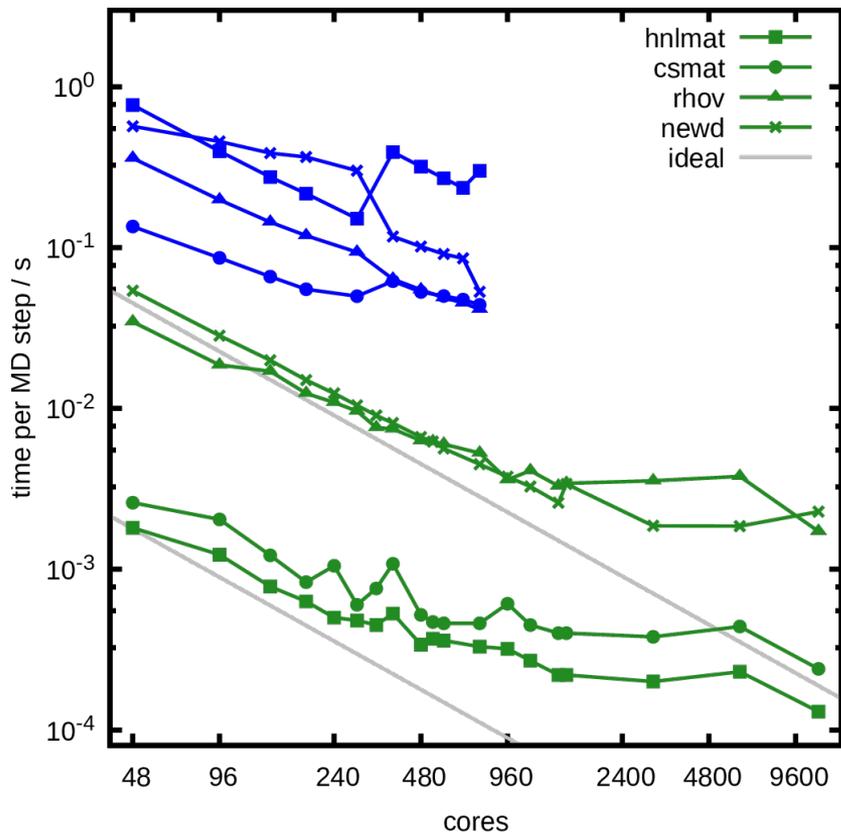
# Calculation of Augmentation Charges & New D

- Save Q-function at initialization
- Parallelization and optimization of becsun calculation
- Blocking of  $N_{hg}$



- Rhov  
DGEMM for each species (2)
- Newd  
Merge DGEMM if ionic forces are needed (2)  
Summation of packed array

# Hnmat & Csmat



- Sophisticated loopnest (7)  
-> loopnest (6) + DGEMM
- Reworked MPI parallelization