

Why PaRSEC is the right runtime for exascale computing

Anthony Danalis

Innovative Computing Laboratory

University of Tennessee

CCDSC'14, Chateau des Contes

George Bosilca
Aurelien Bouteiller
Mathieu Faverge
Thomas Herault
Jack Dongarra



Programming Paradigms

- ✓ PaRSEC offers a new prog. paradigm: PTG
 - ✓ PTG: Parameterized Task Graph
- ✓ What is wrong with current prog. paradigms?
- ✓ What is the current programming paradigm?
 - ✓ Pete says MPI & SPMD are not prog. paradigms

Serial Code

```
do i=0, len  
    y(i) = y(i) + a*x(i)  
enddo
```

OpenMP Code

```
#pragma omp parallel  
do i=0, len  
    y(i) = y(i) + a*x(i)  
enddo
```

MPI Code

```
// ugly code here  
do i=my_start, my_len  
    y(i) = y(i) + a*x(i)  
enddo  
// more ugly code
```

MPI Code

```
// ugly code here  
do i=my_start, my_len  
    y(i) = y(i) + c(i)  
enddo  
// more ugly code
```

Data Distribution, parallelism and load balance are coupled.

MPI Code

```
// ugly code here  
do i = start, my_len  
    y(i) = x(i) + 1  
enddo  
// more ugly
```

No mechanism to deal with jitter.
Data Distribution, parallelism and load balance are coupled.

MPI+X Code

- Yeah, right.

MPI Programs are ...

Coarse Grain Parallelism with Explicit Message Passing

Programs are essentially sequential, maybe with some code to coordinate.

- Vaidy Sunderam

Tile Algorithms

PARALLEL LINEAR ALGEBRA SOFTWARE FOR MULTICORE ARCHITECTURES

PLASMA

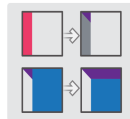
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THE PARALLEL LINEAR ALGEBRA SOFTWARE FOR MULTICORE ARCHITECTURES (PLASMA) PROJECT aims to address the critical and highly disruptive situation that is facing the Linear Algebra and High Performance Computing community due to the introduction of multicore architectures. PLASMA's ultimate goal is to create software frameworks that enable programmers to simplify the process of developing applications that can achieve both high performance and portability across a range of new architectures. PLASMA uses a programming model that allows asynchronous, out-of-order scheduling of operations in order to achieve a scalable yet highly efficient software framework for Computational Linear Algebra applications.

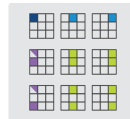
TILE ALGORITHMS

Unlike LAPACK, which uses block algorithms, PLASMA relies on tile algorithms to enable the use of fine grained parallelism.

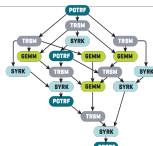
LAPACK
Block algorithms



PLASMA
Tile algorithms



Tile algorithms of Linear Algebra operations can be represented as Directed Acyclic Graphs (DAG) where nodes represent the tasks in which the operation can be decomposed and the edges represent the dependencies among them. As long as the task execution order does not violate the dependencies, the result will be correct.



Example of a DAG for a Cholesky Factorization

PLASMA 2.1.0

- Solution of Linear Equations
- Linear Least Squares Problems
- Multiple Precision Support
- Mixed-Precision Iterative Solver
- Static Scheduling
- LAPACK Interface / Native Interface
- LAPACK-Compliant Error Handling
- LAPACK-Derived Testing Suite
- Thread Safety
- Windows, Linux, AIX, Mac OS
- PLASMA Users' Guide

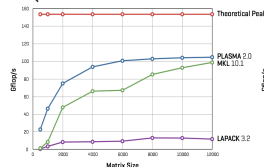
CURRENT RESEARCH

- Singular Value Decomposition
- Symmetric and Non Symmetric Eigenvalue Problems
- Dynamic Scheduling
- Communication Avoiding Algorithms
- Autotuning
- Distributed Memory Machines
- Hardware Accelerators

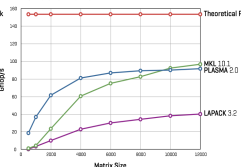
PERFORMANCE RESULTS DOUBLE PRECISION

CPU Intel Xeon 2.4 GHz Quad-socket Quad cores (16 cores total)

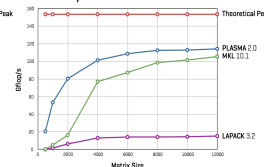
QR Factorization



LU Factorization



Cholesky Factorization



DOWNLOAD THE LIBRARY AT <http://icl.eecs.utk.edu/plasma/>

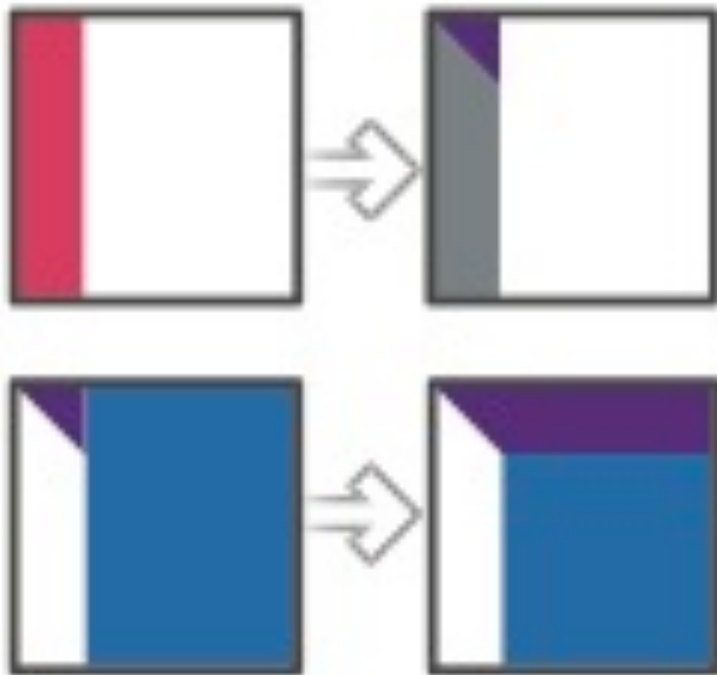
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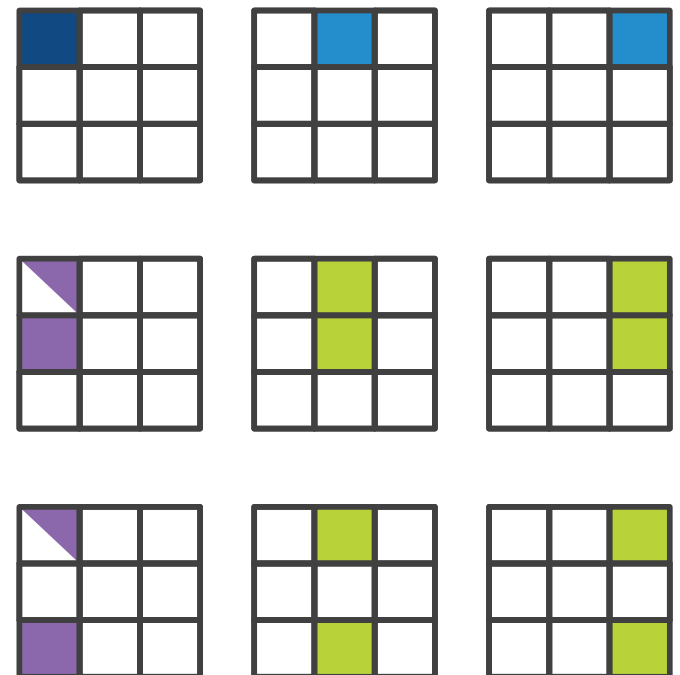
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Panel vs Tile Algorithms

LAPACK / Panel



PLASMA / Tile



What does the code look like?

```
for (k = 0; k < MT; k++) {  
    Insert_Task( zgeqrt,  A[k][k], INOUT,  T[k][k], OUTPUT);  
    for (m = k+1; m < MT; m++) {  
        Insert_Task( ztsqrt,  A[k][k], INOUT | REGION_D|REGION_U,  
                    A[m][k], INOUT | LOCALITY,  
                    T[m][k], OUTPUT);  
    }  
    for (n = k+1; n < NT; n++) {  
        Insert_Task( zunmqr,  A[k][k], INPUT | REGION_L,  
                    T[k][k], INPUT,  
                    A[k][m], INOUT);  
        for (m = k+1; m < MT; m++) {  
            Insert_Task( ztsmqr, A[k][n], INOUT,  
                        A[m][n], INOUT | LOCALITY,  
                        A[m][k], INPUT,  
                        T[m][k], INPUT);  
        }  
    }  
}
```

What's wrong with this code

✗ It has:

- ✗ Control Flow
- ✗ Hints for runtime to infer Data Flow
- ✗ High memory requirements or reduced parallelism

✓ It should have:

- ✓ No (or minimal, user defined) Control Flow
- ✓ Explicit Data Flow
- ✓ Unhindered parallelism

Parameterized Task Graph

- Compressed form of the Execution DAG
- Fixed size (problem size independent)
- Task Classes w/ parameters
 - $\text{geqrt}(k)$, $\text{tsqrt}(k,m)$, $\text{unmqr}(k,n)$, $\text{tsmqr}(k,n,m)$
- Precedence constraints between Tasks

PTG: PING-PONG

```
PING(s)
  s = 0..max_steps-1
  : A(s)
  RW    A0 <- A(s)
        -> A0 PONG(s)
  READ  A1 <- (s != 0) ? PONG(s-1)
  BODY  verify_response(A0, A1); END
```

```
PONG(s)
  s = 0..max_steps-2
  : A(s+1)
  RW    A0 <- A0 PING(s)
        -> A1 PING(s+1)
  BODY  /* do nothing on data */ END
```

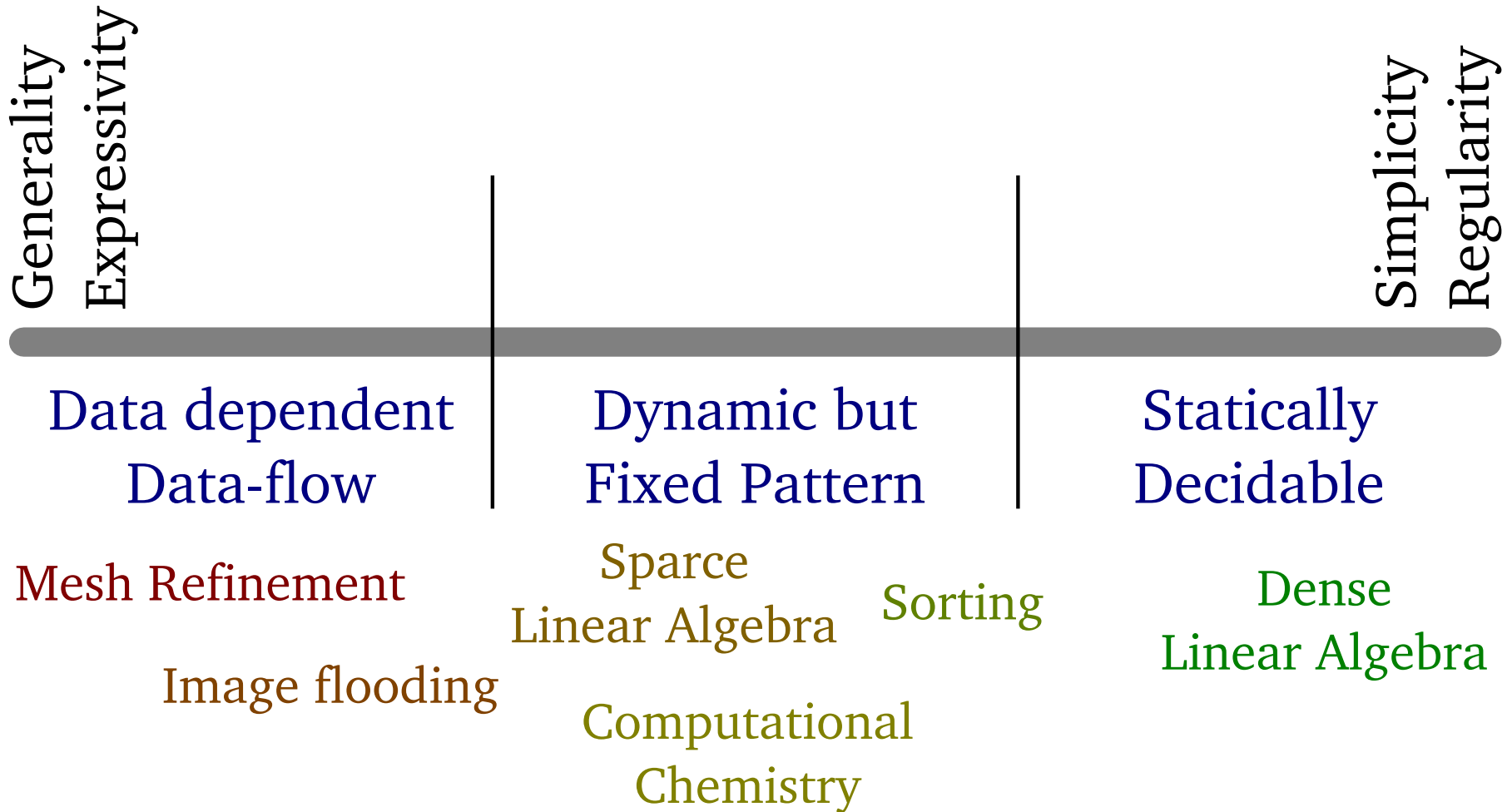
PTG: Binary Tree Reduction

```
BT_REDUC(tree, step, i)
  tree_count = count_bits(NT)
  tree = 1 .. tree_count
  max_step = log_of_tree_size(NT, tree)
  step = 1 .. max_step
  i = 0 .. (1<<(max_step-step))-1
  offset = compute_offset(NT, tree)

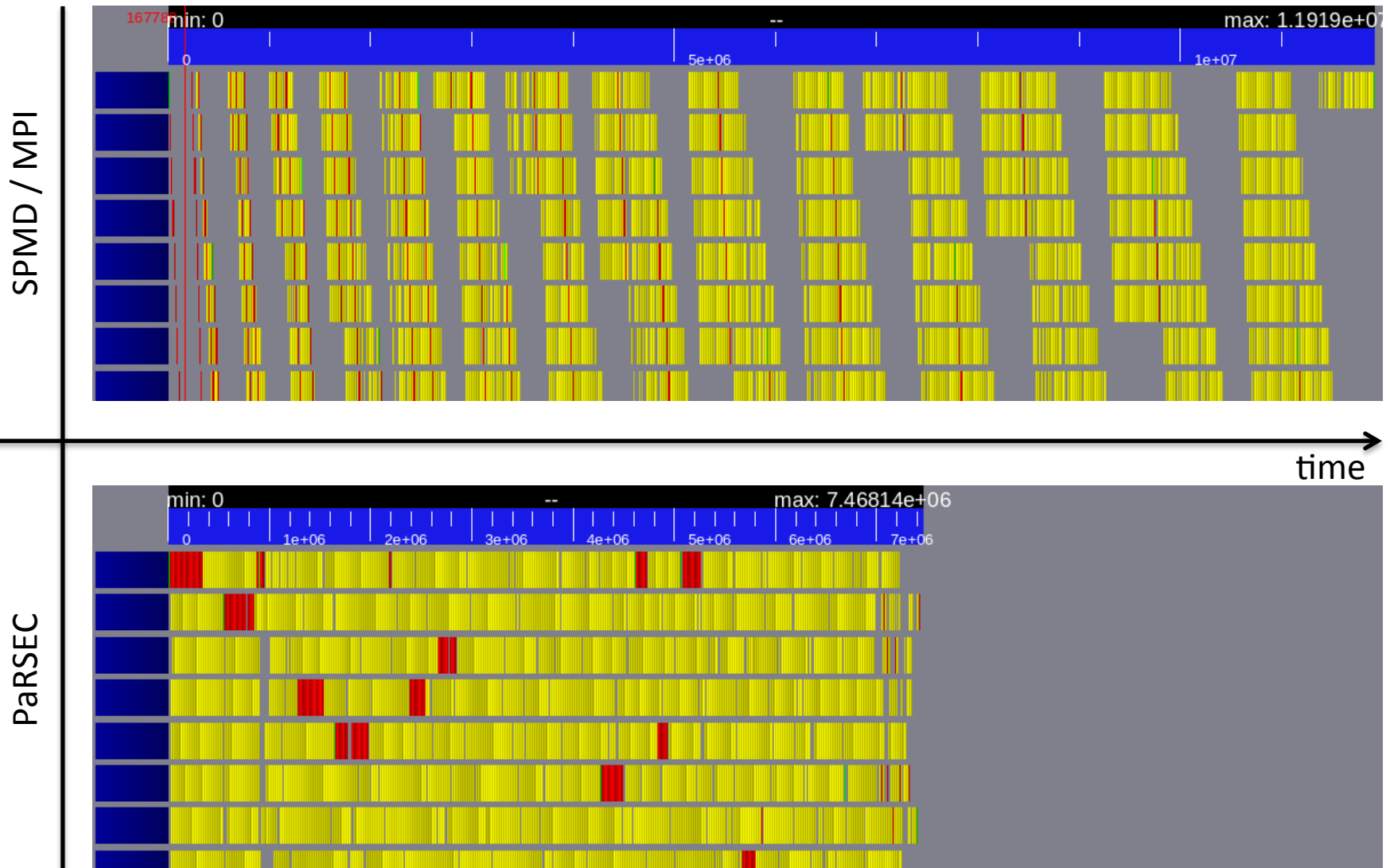
  : dataA(offset+i*2,0)

READ A  <- (1==step) ? A REDUCTION(offset+i*2)
        <- (1!=step) ? B BT_REDUC(tree, step-1, i*2)
RW      B  <- (1==step) ? A REDUCTION(offset+i*2+1)
           <- (1!=step) ? B BT_REDUC(tree, step-1, i*2+1)
           -> ((max_step!=step) && (0==i%2)) ? A BT_REDUC(tree, step+1, i/2)
           -> ((max_step!=step) && (0!=i%2)) ? B BT_REDUC(tree, step+1, i/2)
           -> (max_step==step) ? C LINEAR_REDUC(tree)
BODY int j; for(j=0; j<NB; j++){ REDUCE( A, B, j ); } END
```

Who is this for?



Load Balance, Idle time & Jitter

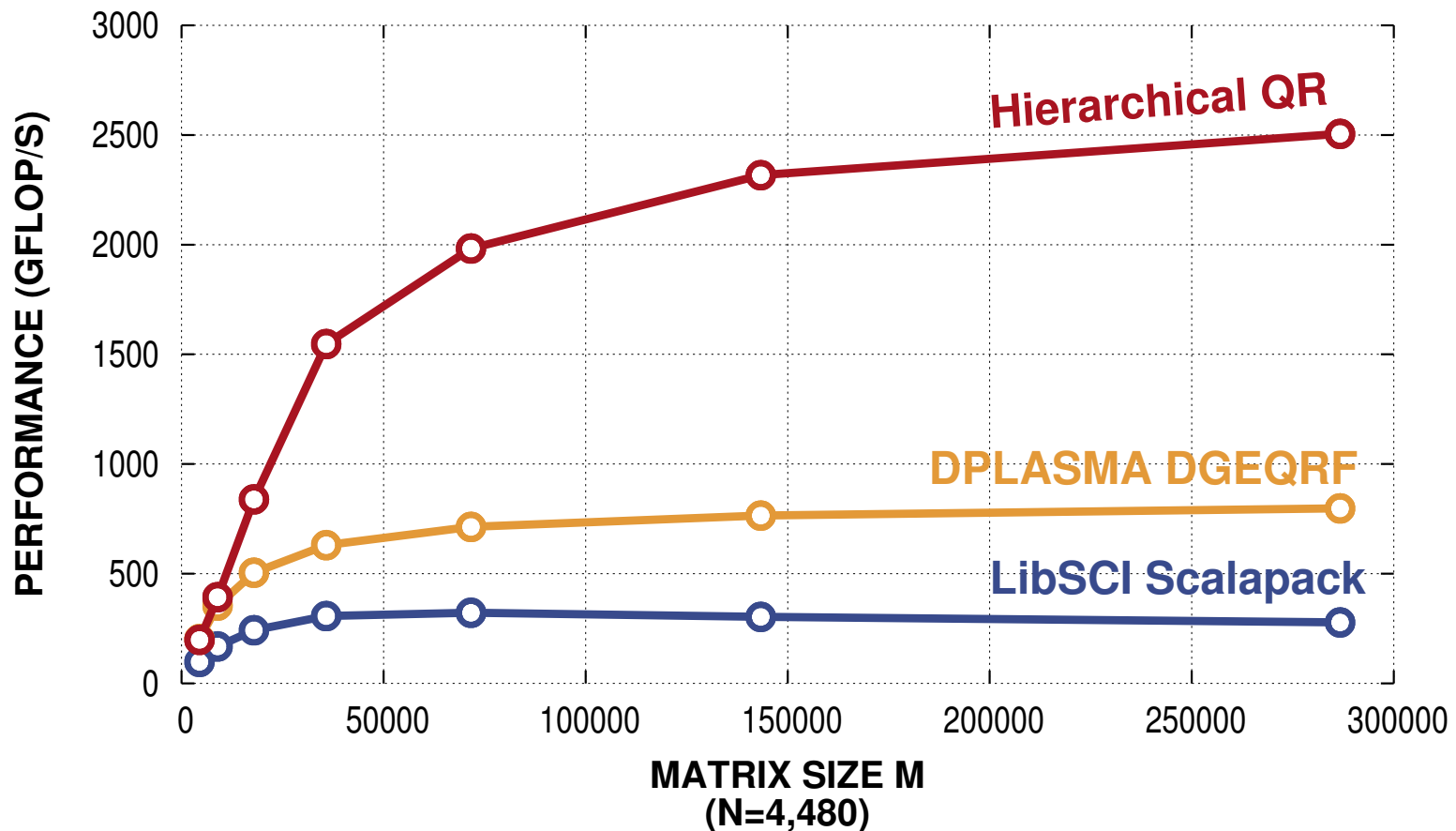


Performance: (H)QR

Solving Linear Least Square Problem (DGEQRF)

60-node, 480-core, 2.27GHz Intel Xeon Nehalem, IB 20G System

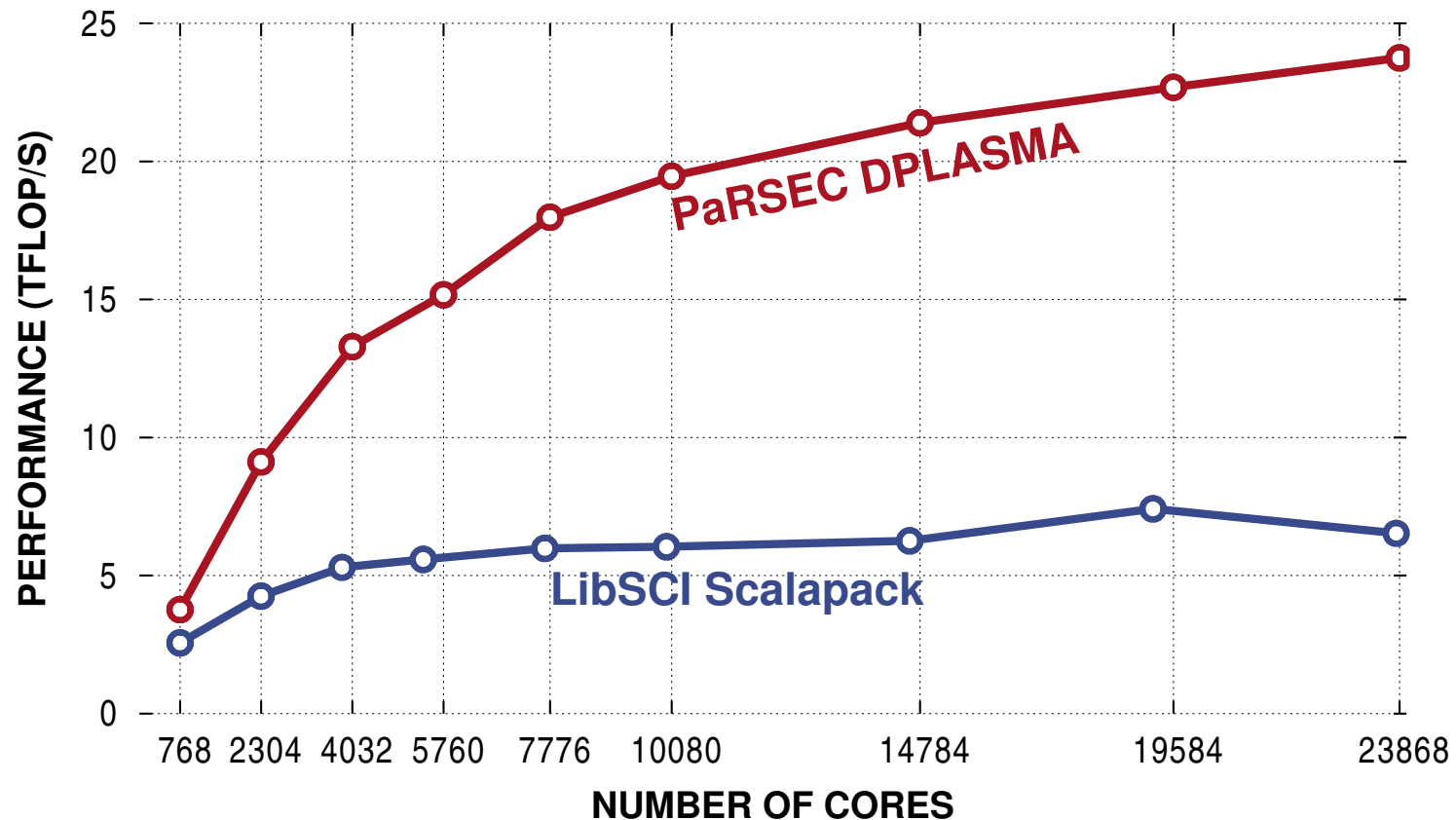
Theoretical Peak: 4358.4 GFlop/s



Performance: Systolic QR

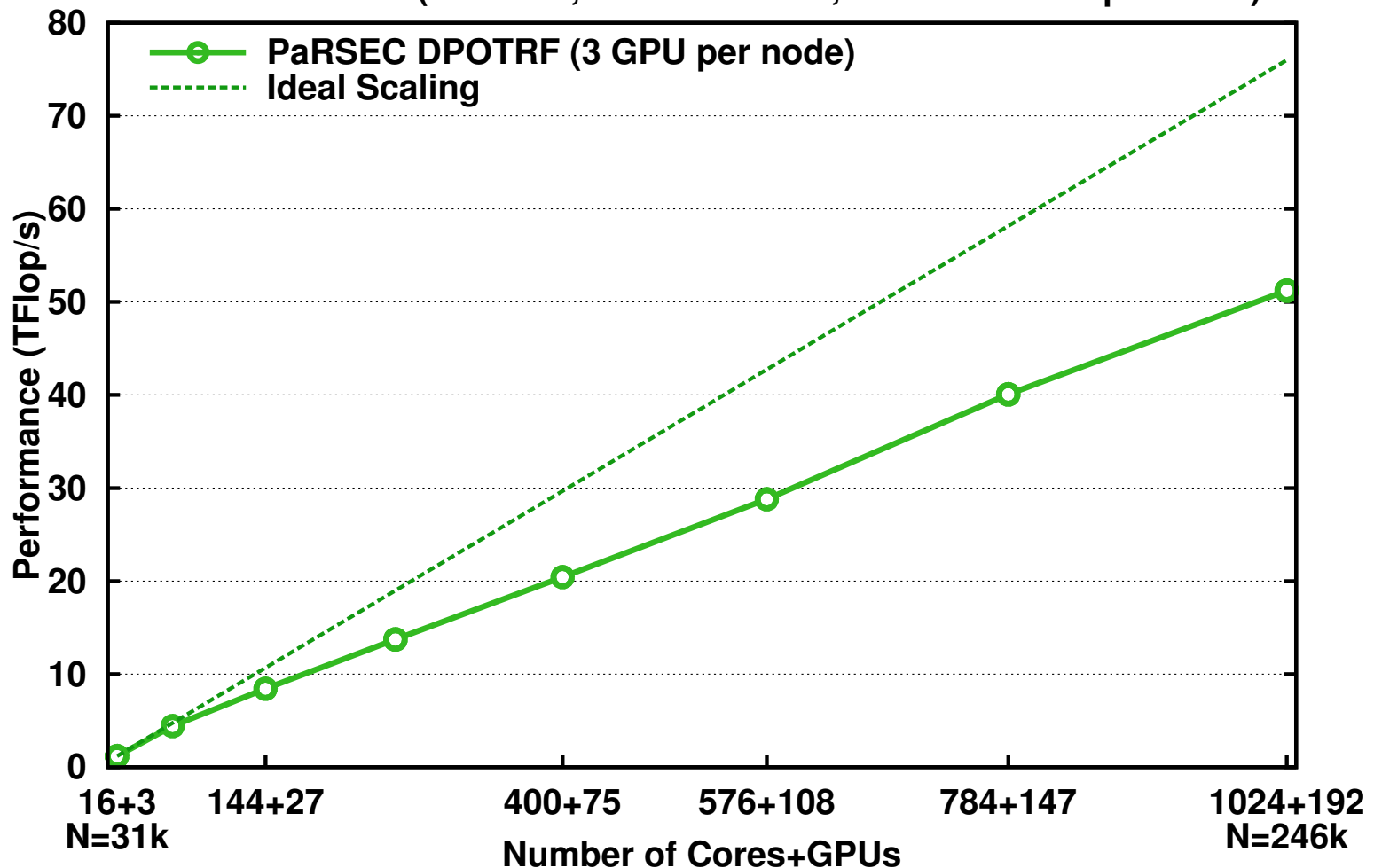
DGEQRF performance strong scaling

Cray XT5 (Kraken) - $N = M = 41,472$



Distributed CPUs + GPUs

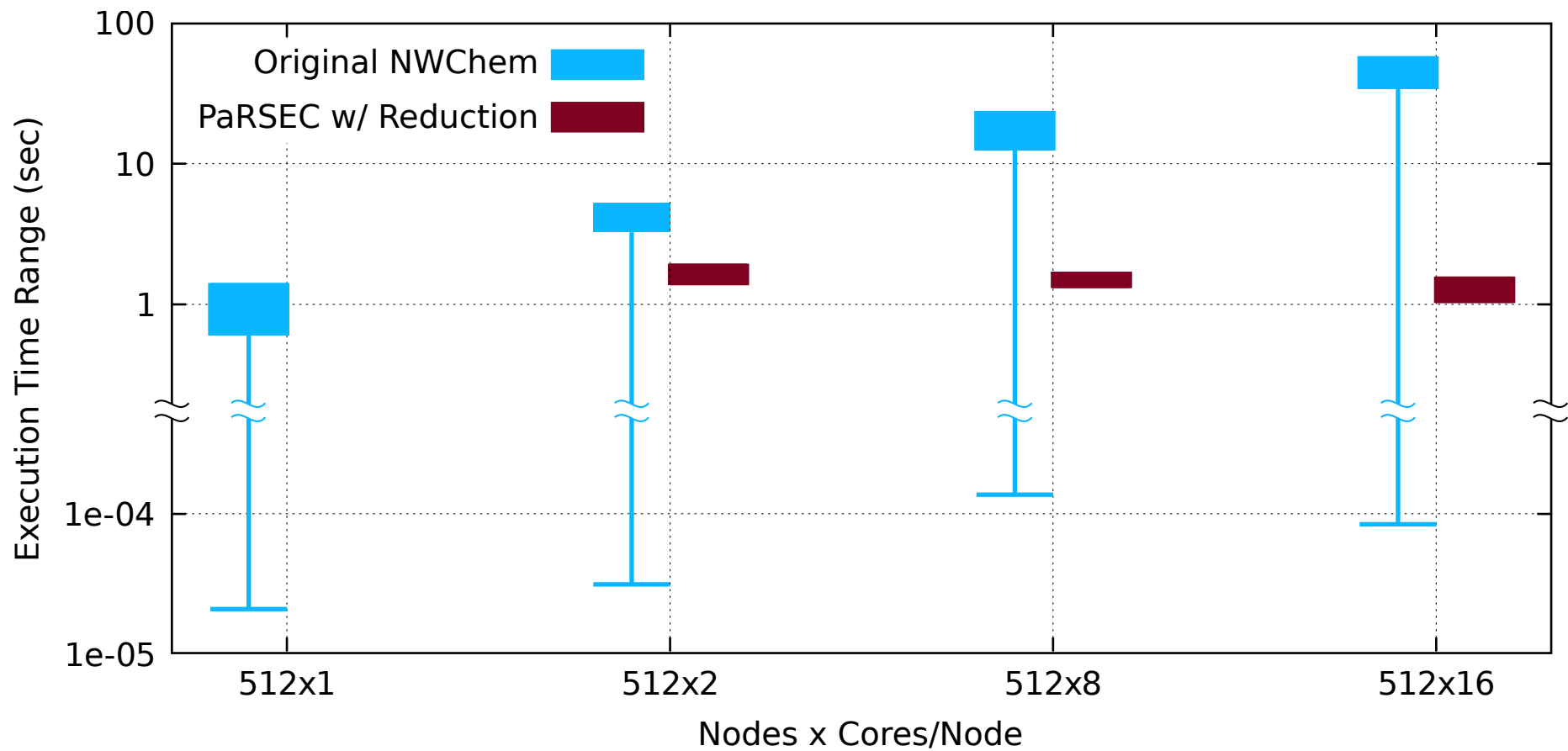
Distributed Hybrid DPOTRF Weak Scaling on Keeneland
1 to 64 nodes (16 cores, 3 M2090 GPUs, Infiniband 20G per node)



NWChem Coupled Cluster (CC)

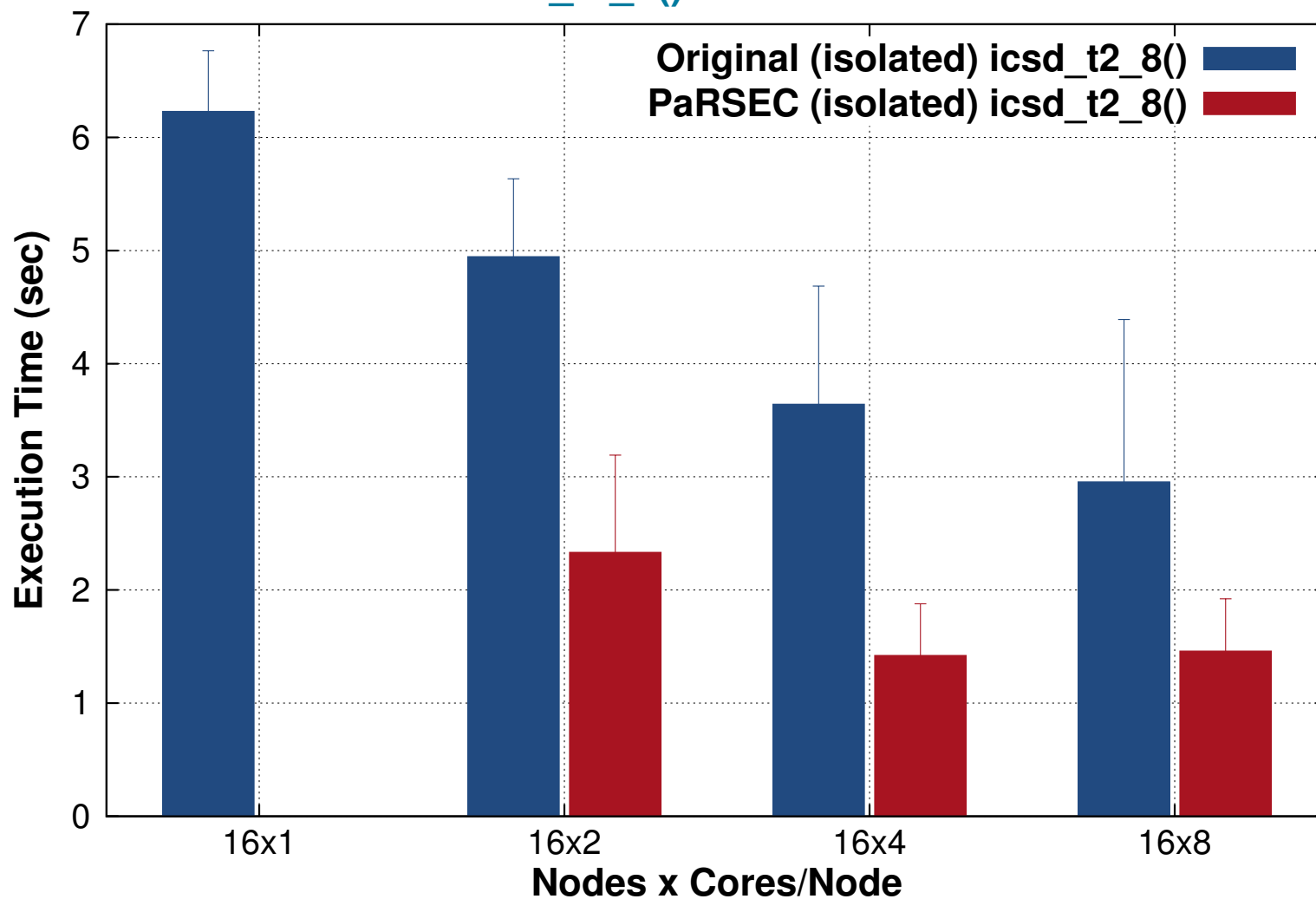
- Computational Chemistry
- TCE Coupled Cluster
- Machine generated (awful) Fortran 77 code
- Behavior depends on dynamic, **immutable** data
- Long sequential chains of DGEMMs
- Work (chain) stealing through GA atomics
- Did I mention that the chains are sequential?

CC t1_2_2_2



CC t2_8

Execution Time of icسد_t2_8() subroutine in CCSD of NWChem



Breaking the chains

```
DO p3b, p4b, h1b, h2b
```

```
    CALL DFILL(dimc,0.0d0,dbl_mb(k_c_sort),1)
```

```
    DO p5b, p6b
```

```
        IF (int_mb(k_spin+p5b-1) .eq. ...) THEN
```

```
            CALL DGEMM( ... )
```

```
        END IF
```

```
    END DO
```

```
CALL TCE_SORT_4(dbl_mb(k_c_sort),dbl_mb(k_c), ... )
```

```
CALL ADD_HASH_BLOCK(d_c,dbl_mb(k_c),dimc, expr )
```

```
END DO
```

Breaking the chains

```
DO p3b, p4b, h1b, h2b
```

```
CALL DFILL(dimc,0.0d0,dbl_mb(k_c_sort),1)
```

```
DO
```

```
IF (int_mb(k_spin+p5b-1) .eq. ...) THEN
```

```
C = C + A*B
```

```
ENDDO
```

```
CALL TCE_SORT_4(dbl_mb(k_c_sort),dbl_mb(k_c), ... )
```

```
CALL ADD_HASH_BLOCK(d_c,dbl_mb(k_c),dimc, expr )
```

```
END DO
```

Breaking the chains

```
DO p3b, p4b, h1b, h2b
```

```
    CALL DFILL(dimc,0.0d0,dbl_mb(k_c_sort),1)
```

```
DOANY
```

```
    IF (int_mb(k_spin+p5b-1) .eq. ...) THEN
```

```
        C = C + A*B
```

```
    END IF
```

```
ENDDO
```

```
CALL TCE_SORT_4(dbl_mb(k_c_sort),dbl_mb(k_c), ... )
```

```
CALL ADD_HASH_BLOCK(d_c,dbl_mb(k_c),dimc, expr )
```

```
END DO
```

Conclusions

- PTG offers a Data-Flow based Prog. Paradigm
- PaRSEC offers state-of-the-art performance
- Data distribution is decoupled from Algorithm
- And you can bring your GPUs too
- And you can even bring your real applications
- CGP (MPI/SPMD) != highest performance
- CGP != easiest to program (?)