



# XKaapi : a runtime for highly parallel (OpenMP) application

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MOAIS, INRIA, Grenoble

# Agenda

- context
- objective
- task model and execution
- status

# Computer



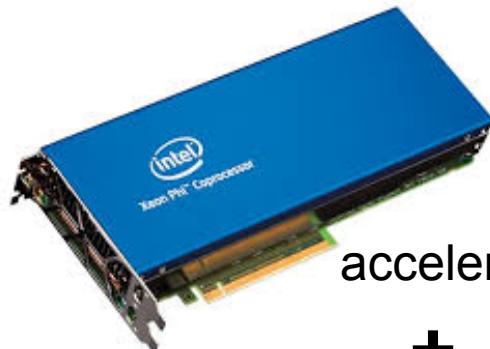
processor

+



memory

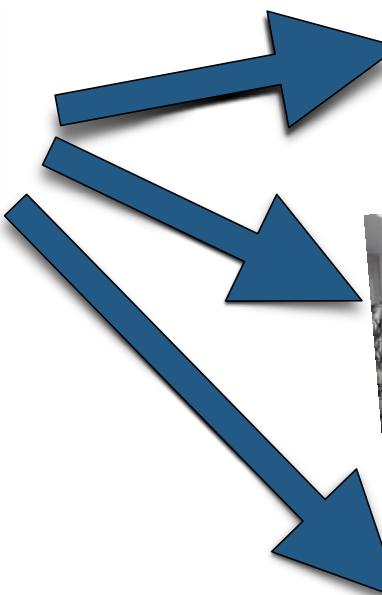
+



accelerator



+



# Processor architecture



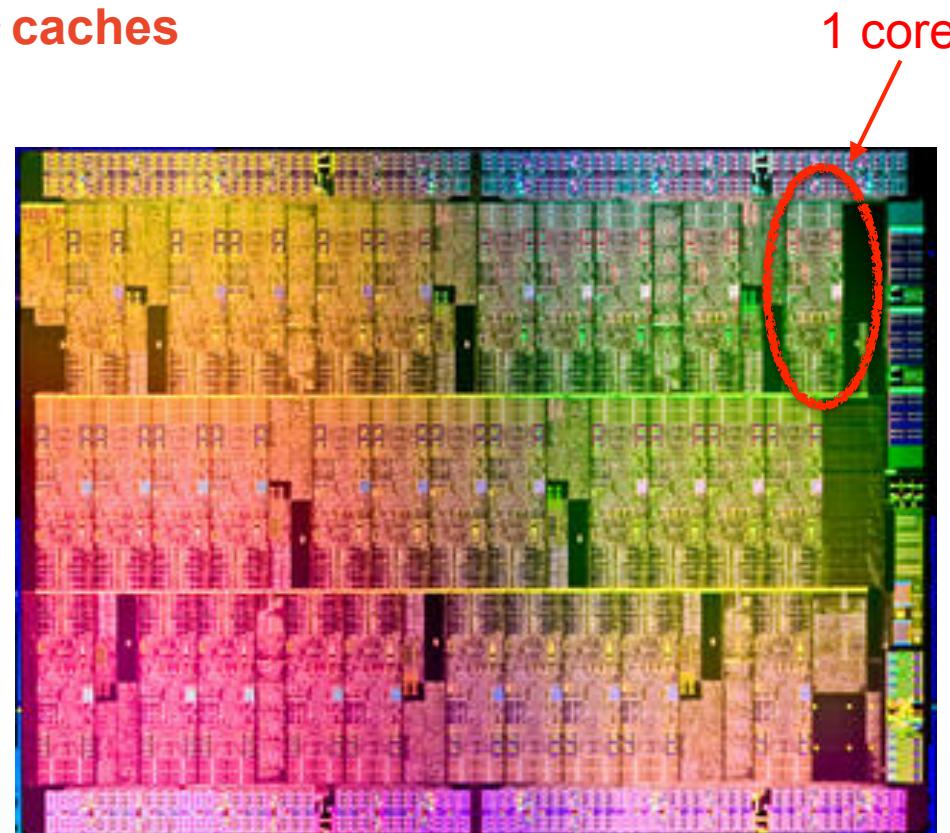
- **multicore**
- **caches**



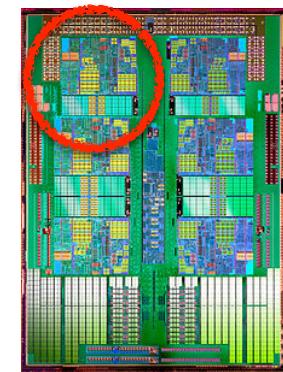
4-cores AMD Barcelona



4-cores Haswell

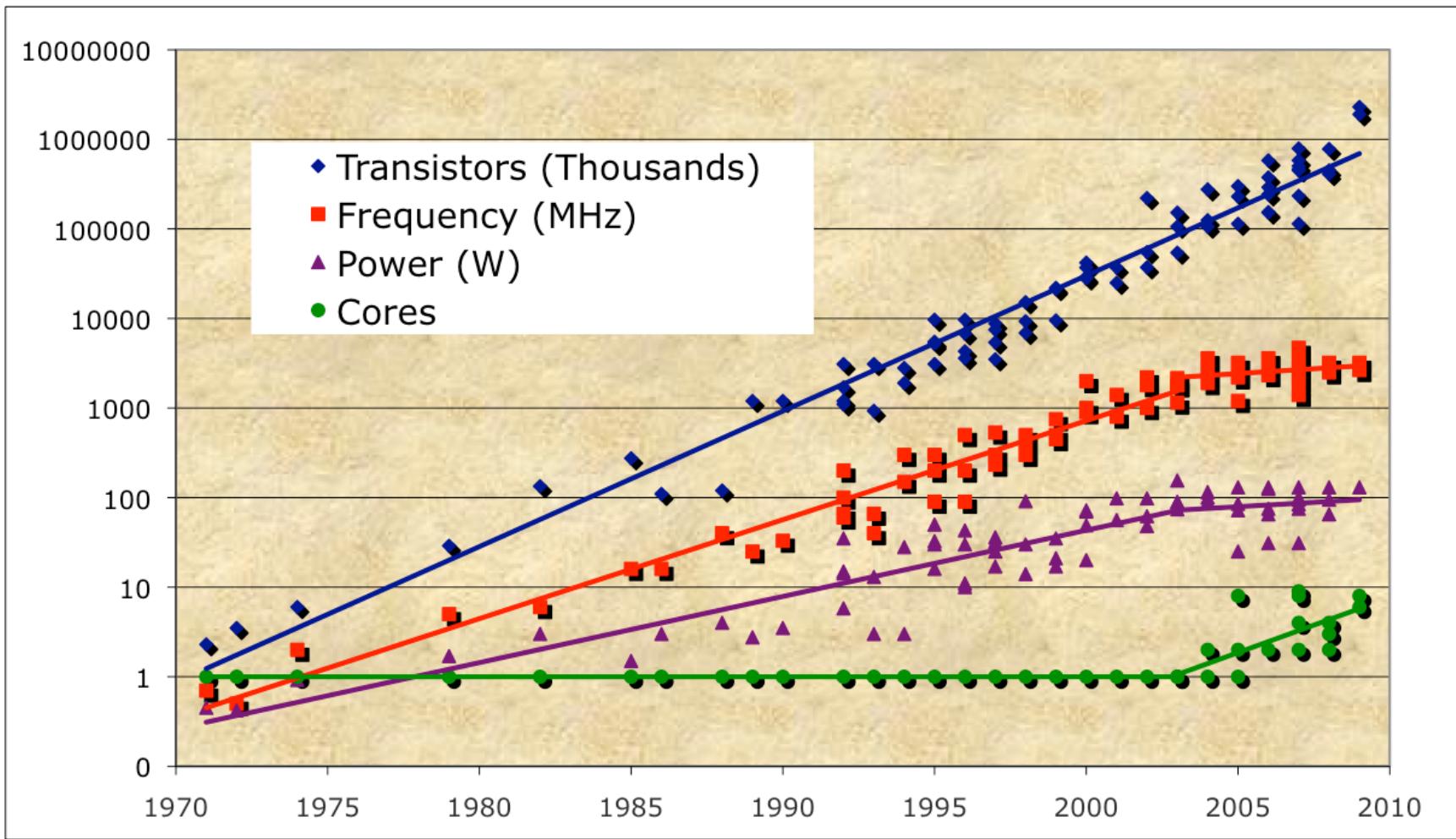


Intel Xeon Phi, 60 cores !



6-cores AMD Istanbul

# Trends



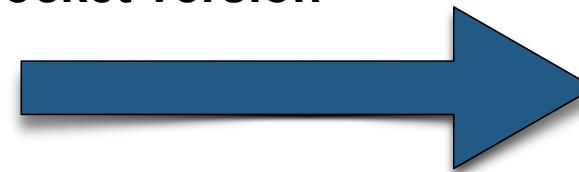
**Data from** Kunle Olukotun, Lance Hammond, Herb Sutter, Burton Smith, Chris Batten, and Krste Asanović  
**Slide from** Kathy Yellick

- many-many cores
- cache hierarchies
- hybrid
  - AMD Fusion, Nvidia

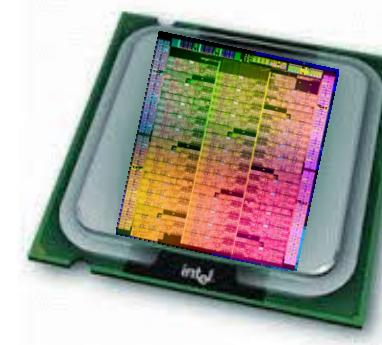
# Knights Landing: multi-many cores



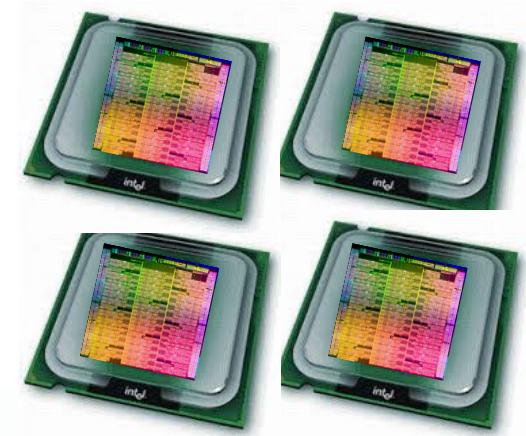
~ 2015  
socket version



No more offloading  
Direct access to main memory



- 4 sockets Intel Xeon Phi machine ?
  - $\sim 4 * 72 = 288$  cores
  - $\sim 4 * 4 * 72 = 1152$  hyperthreads
  - $\sim 4 * 3$  Tflop/s
  - what about memory coherency... ?



# How to program them ?

- **Top-down classification**

- ▶ **network**

- MPI, PGAS (UPC), X10

- ▶ **multi-core**

- OpenMP, X10, Cilk, Intel TBB, XKaapi, StarPU...

- ▶ **accelerator**

- Cuda, OpenCL, OpenACC, OpenMP (4.0)

- ▶ **vecteur / unité SIMD**

- compilateur, type étendu, OpenMP (4.0)



OpenMP 4.0 !

- **Which parallel programming environment ?**

- ▶ **OpenMP-4.0**

- + OpenCL?
    - + Cuda?

- **Challenge : same programming model, same code.**

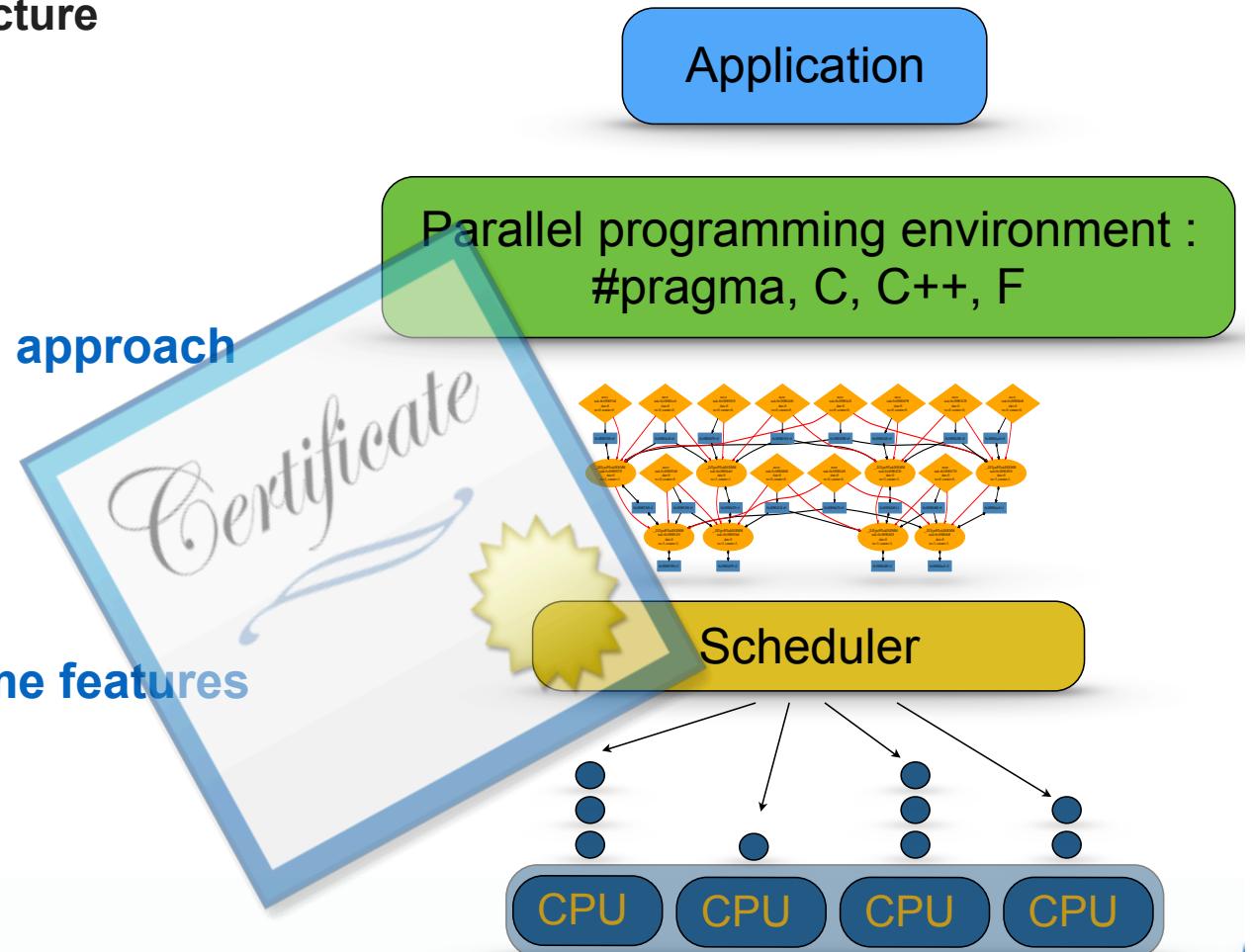
# Challenge: performance portability!

- Performance portability
  - on several generations of architecture
  - with load balancing issues
    - OS jitter, irregular application

- MOAIS team promotes a 2 steps approach

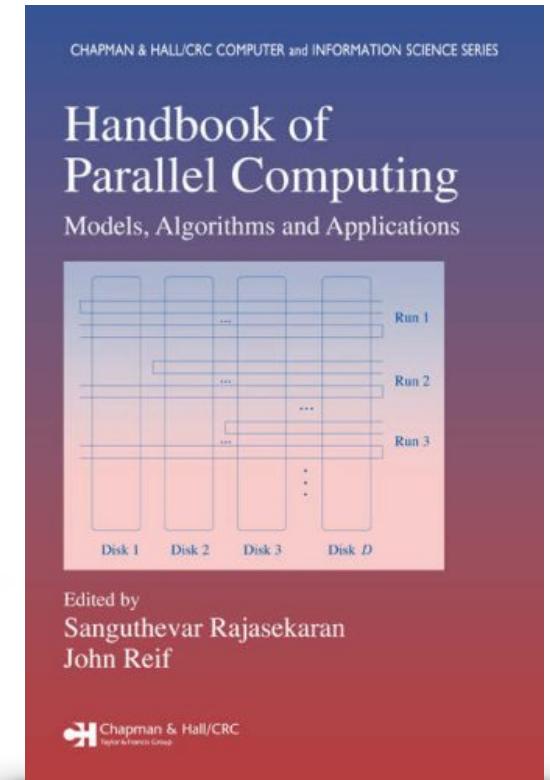
1. parallel algorithms
  - communications !
2. scheduling as a plugin

- Requirement of high level runtime features



# Which runtime features ?

- Programming model pressure
    - Task (independent and dependent)
      - extension
    - Parallel loop
  - Architecture pressure
    - High degree of parallelism
    - Memory hierarchy
  - Performance guarantee
    - parallel algorithms
    - scheduling
      - work stealing, HEFT, Dual approximation
      - ...
- Management of many-(many) tasks
- parallel slackness
  - spatial and temporal locality
  - scalable internal algorithms and data structure



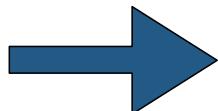
# XKaapi task model

- Athapaskan task model
  - description of data dependencies (read/write/accumulation)
- Using OpenMP code annotation

```
void main()
{
    /* data result is produced */
    compute( input, &result );

    /* data result is consumed */
    display( &result );

}
```

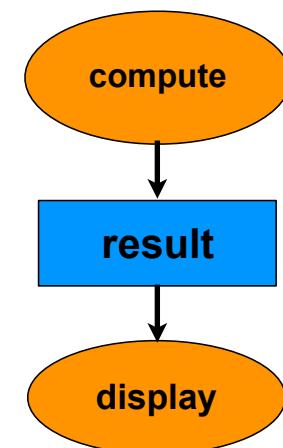


```
void main()
{
    #pragma omp task depend(out: result)
    compute( input, &result );

    #pragma omp task depend(in: result)
    display( &result );

}
```

}



- Task
  - OpenMP structured block, function call
  - assumption: no side effect, description of access mode
- ~ Sequential semantics

# Limitation

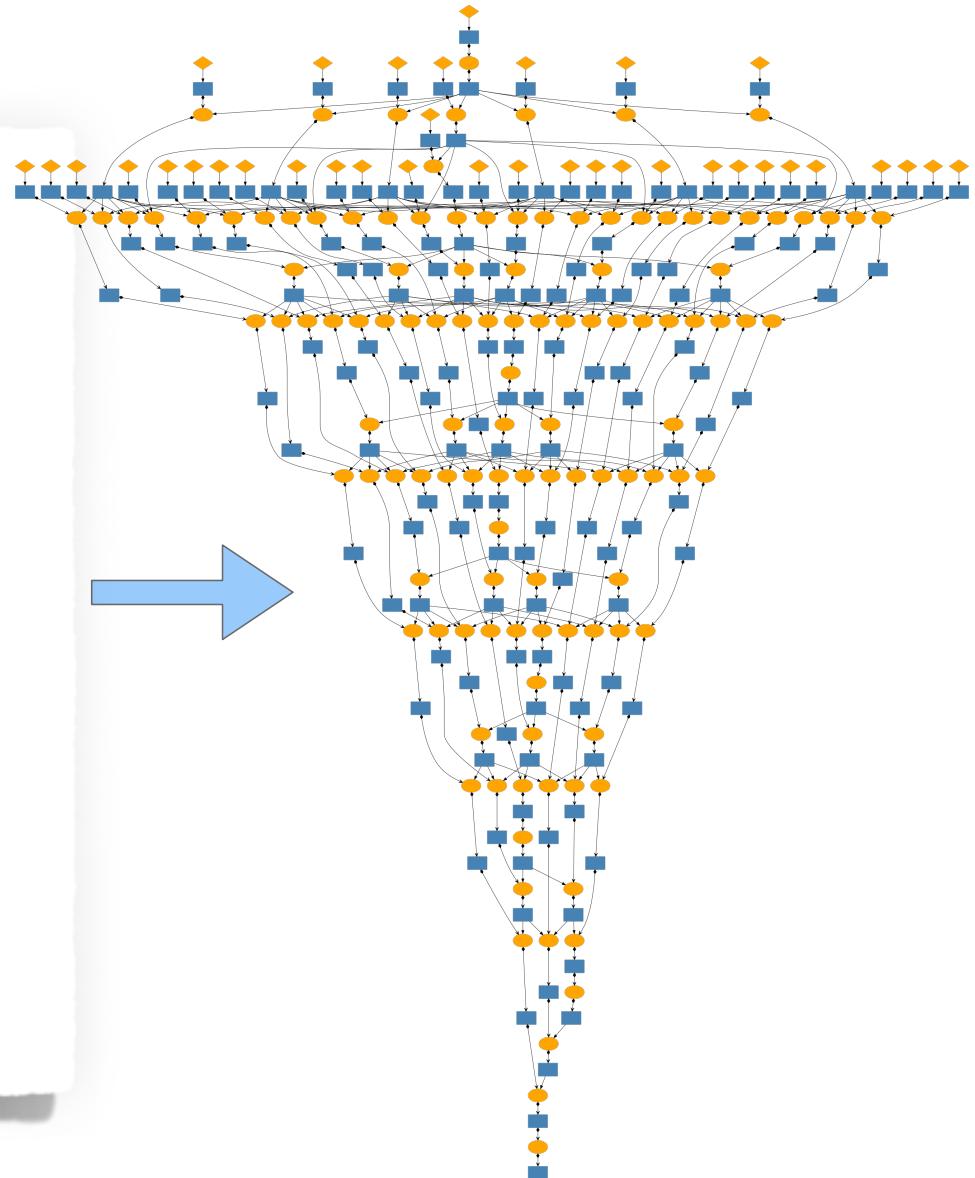
```
#include <cbLAS.h>
#include <clapack.h>

void Cholesky( int N, double A[N][N], size_t NB )
{
    for (size_t k=0; k < N; k += NB)
    {
        #pragma omp task depend(inout: A[k:NB][k:NB]) shared(A)
        clapack_dpotrf( CblasRowMajor, CblasLower, NB, &A[k*N+k], N );

        for (size_t m=k+ NB; m < N; m += NB)
        {
            #pragma omp task depend(in: A[k:NB][k:NB]) \
                depend(inout: A[m:NB][k:NB]) shared(A)
            cblas_dtrsm( CblasRowMajor, CblasLeft, CblasLower, CblasNoTrans, CblasUnit,
                         NB, NB, 1., &A[k*N+k], N, &A[m*N+k], N );
        }

        for (size_t m=k+ NB; m < N; m += NB)
        {
            #pragma omp task depend(in: A[m:NB][k:NB]) \
                depend(inout: A[m:NB][m:NB]) shared(A)
            cblas_dsyrk ( CblasRowMajor, CblasLower, CblasNoTrans,
                          NB, NB, -1.0, &A[m*N+k], N, 1.0, &A[m*N+m], N );

            for (size_t n=k+NB; n < m; n += NB)
            {
                #pragma omp task depend(in: A[m:NB][k:NB], A[n:NB][k:NB])\
                    depend(inout: A[m:NB][n:NB]) shared(A)
                cblas_dgemm ( CblasRowMajor, CblasNoTrans, CblasTrans,
                              NB, NB, NB, -1.0, &A[m*N+k], N, &A[n*N+k], N, 1.0, &A[m*N+n], N );
            }
        }
    }
    #pragma omp taskwait
}
```



- Size of the graph !
- Two complementary approaches
  - fine grain task management: XKaapi task creation ~ 10cycles / task
  - compact dag representation for symbolic scheduling (DAGuE [UTK]), lazy task creation

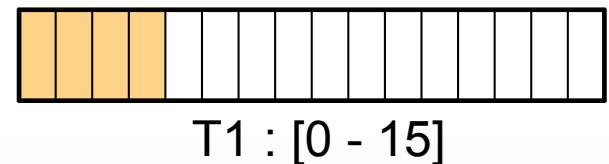
# Adaptive Task model

- **On-demand task creation**

- Adaptive tasks can be split at run time to create new tasks
- Provide a «splitter» function called by idle cores on running task
  - steal part of the remaining computation

- **Typical example : parallel loop**

- A task == a range of iterations
- Initially, one task in charge of the whole range



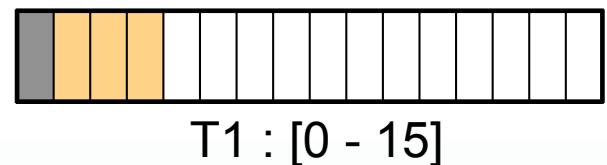
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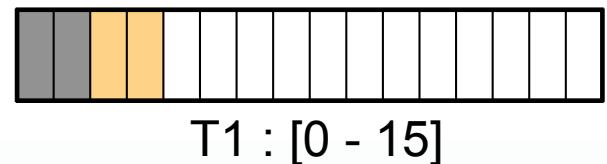
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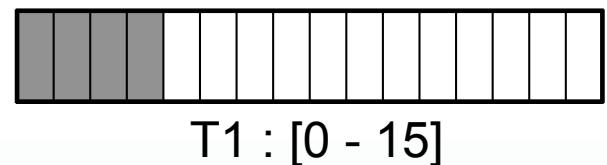
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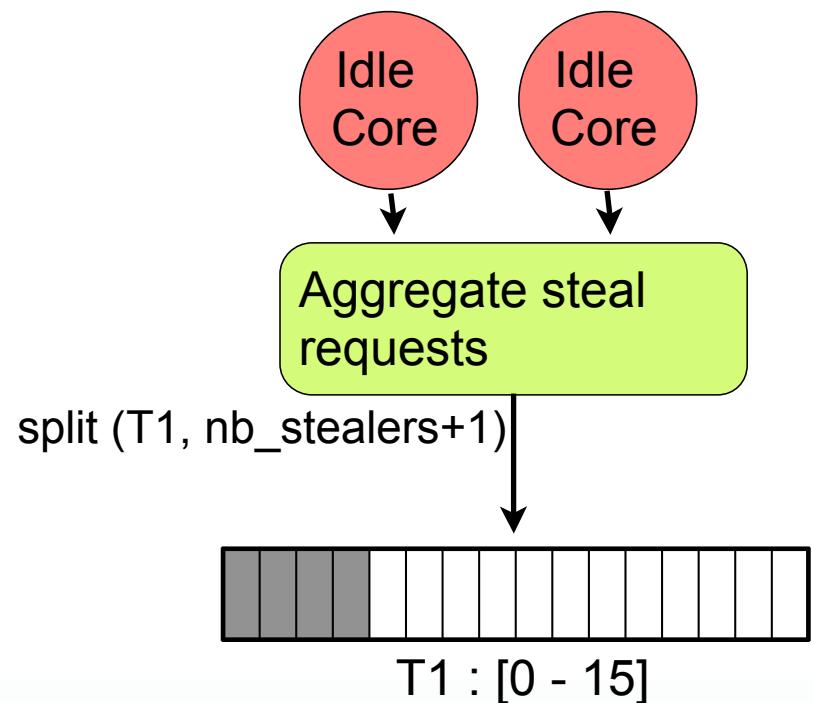
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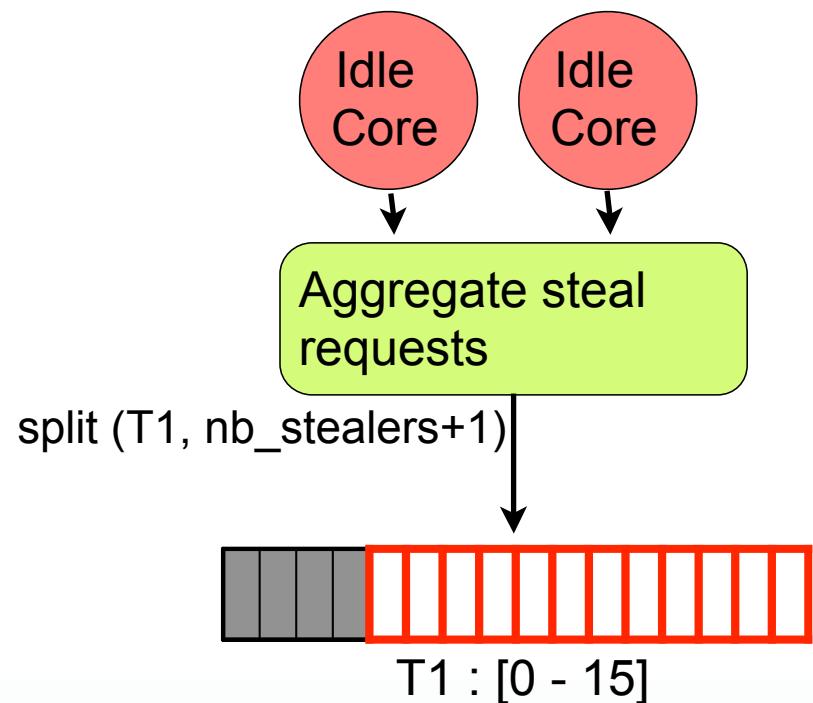
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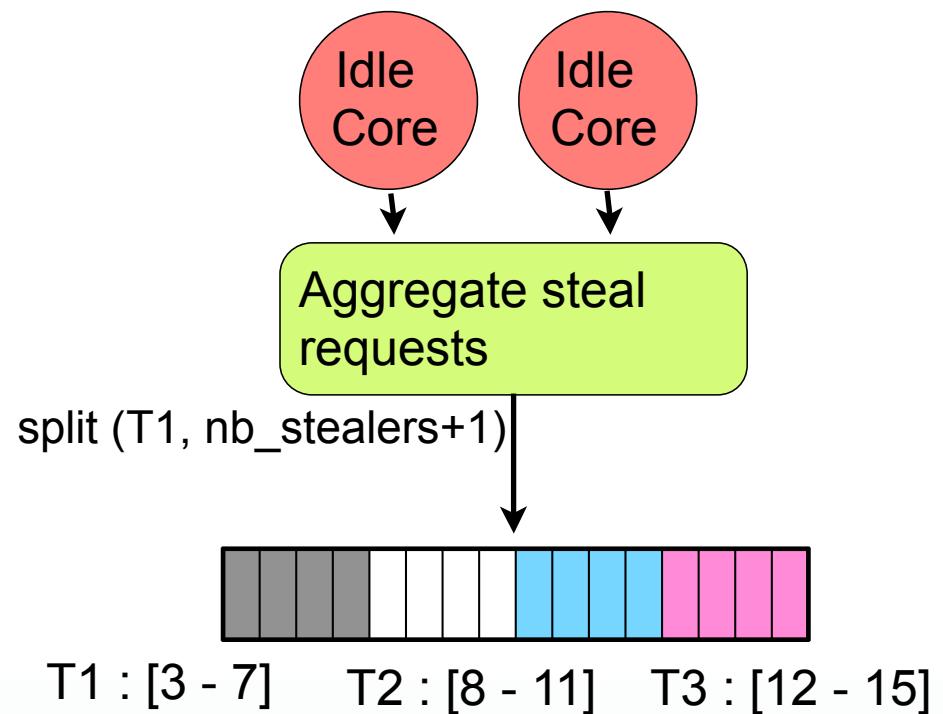
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# The way XKaapi executes tasks

- **Work-stealing based scheduling**

- One “worker thread” per core
    - Holds a queue of tasks

- **push = Task creation is cheap !**

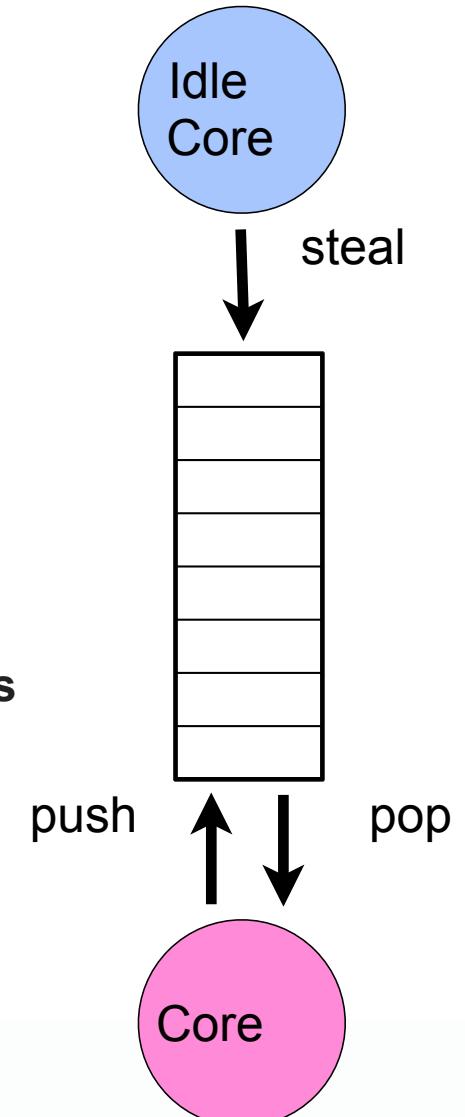
- Reduces to pushing C function pointer + its arguments into the worker thread queue
    - ~ 10 cycles / tasks on AMD Many Cours processors
  - Recursive tasks are welcome

- **pop = Task execution by the owner is cheap !**

- pop the next task in the queue without computing dependencies
    - sequential execution is a valid order

- **steal = the most costly operation**

- Compute ready tasks due dependencies
    - Cilks's work first principle !



# Steal operation

- **Theoretical basis**

- Work first principle: if application is highly parallel, steal operations are rare
- Move overhead from the work to the critical path

- **Adaptive task**

- **Detection of ready task**

- visit all tasks in a work queue following sequential order of creation
  - for each task, visit all data accesses to detect dependencies with previous accesses
- detection of dependencies can be cached between steal operations
  - allows to build the data flow graph for HEFT or GDUAL scheduler
  - new implementation in XKaapi-3.0, on going work

- **Basic work stealing protocol**

- idle core locks the victim queue

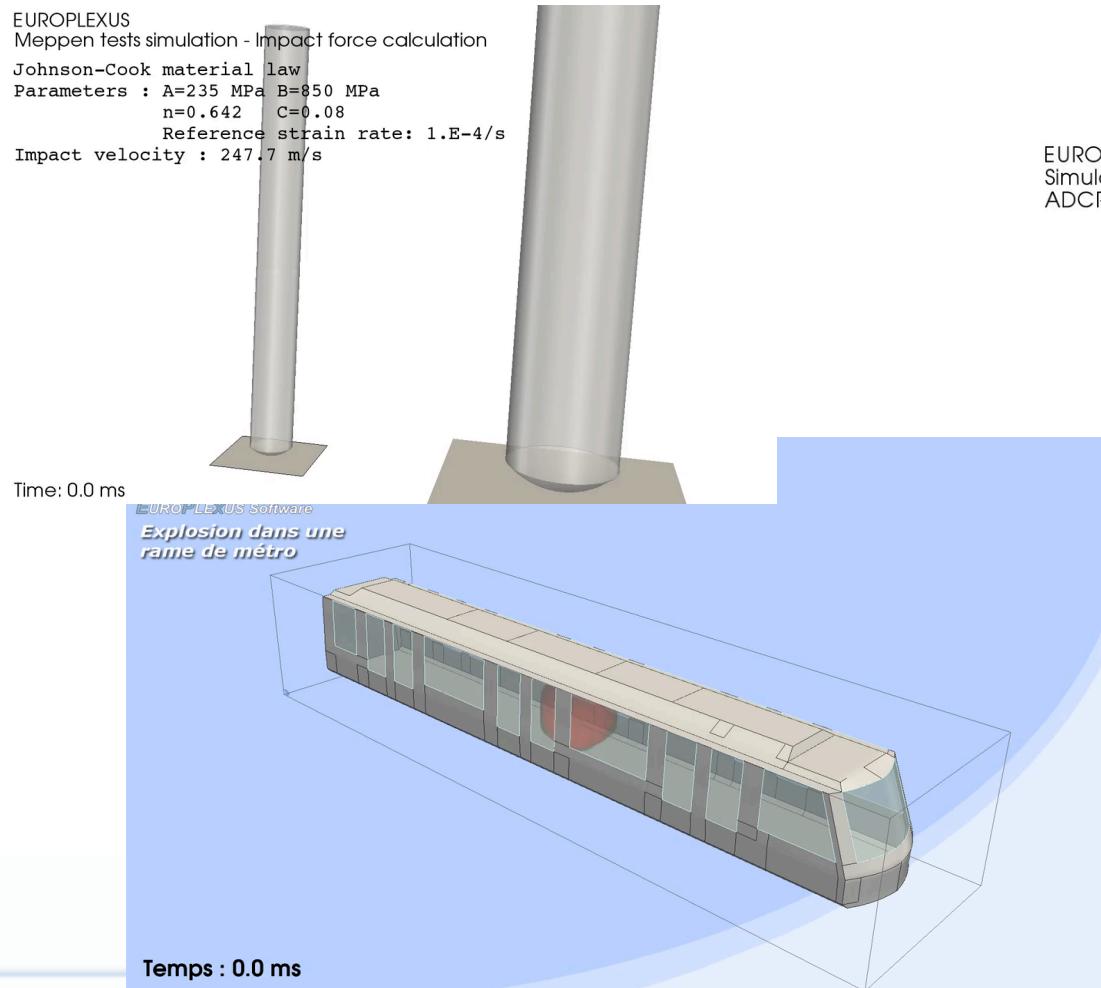
- **Extension : work stealing requests aggregation**

- idle core steals tasks for itself and all waiting cores (on the same victim)
- only one visitor of the victim queue

# Case of study: EPX

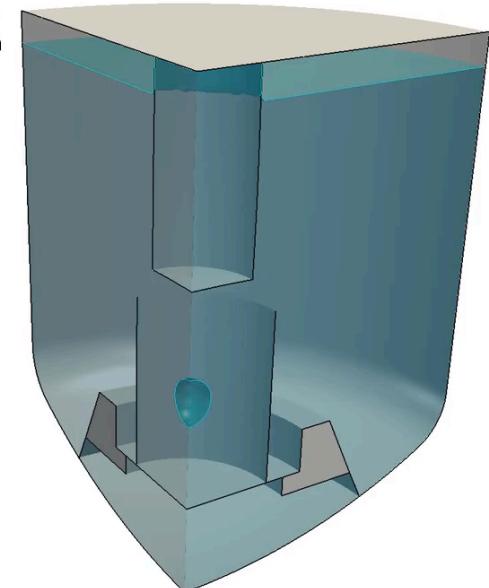
- EPX (EUROPLEXUS) code [CEA - IRC - EDF - ONERA], V. Faucher
  - Fluid-Structure systems subjected to fast transient dynamic loading
- Grand prix SFEN 2013
  - <http://energies.sfen.org/emploi/le-grand-prix-sfen>

EUROPLEXUS  
Meppen tests simulation - Impact force calculation  
Johnson-Cook material law  
Parameters : A=235 MPa B=850 MPa  
 $n=0.642$   $C=0.08$   
Reference strain rate: 1.E-4/s  
Impact velocity : 247.7 m/s



EUROPLEXUS  
Simulation of MARA10 experiment  
ADCR material - VOFIRE algorithm

0.0 ms

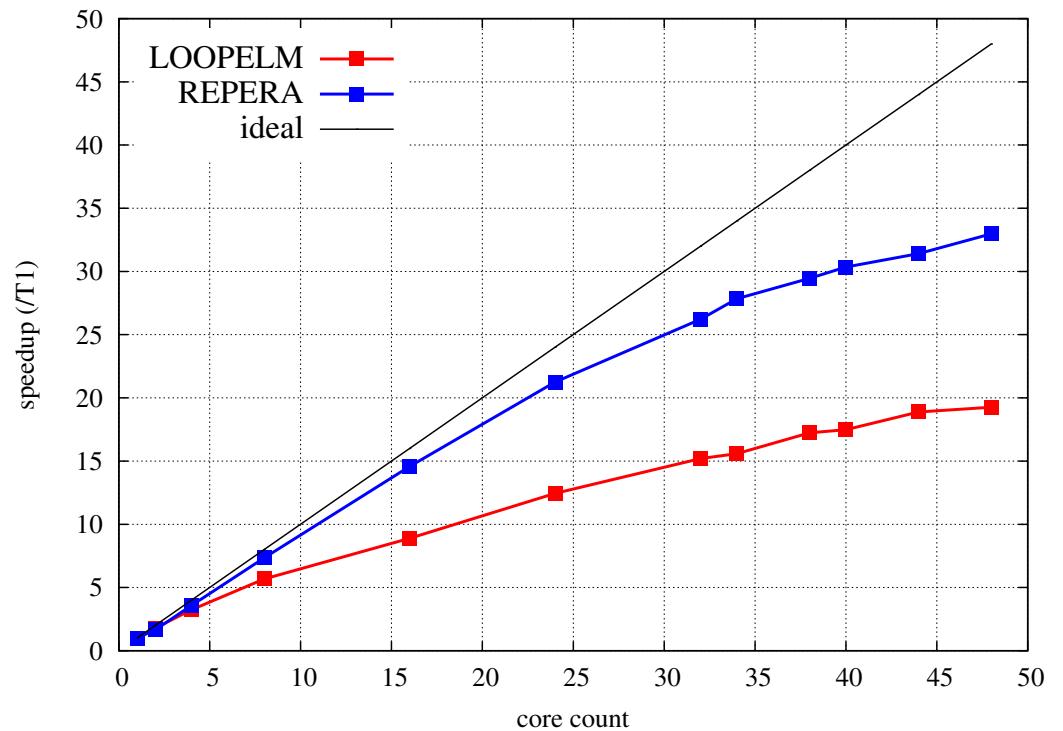
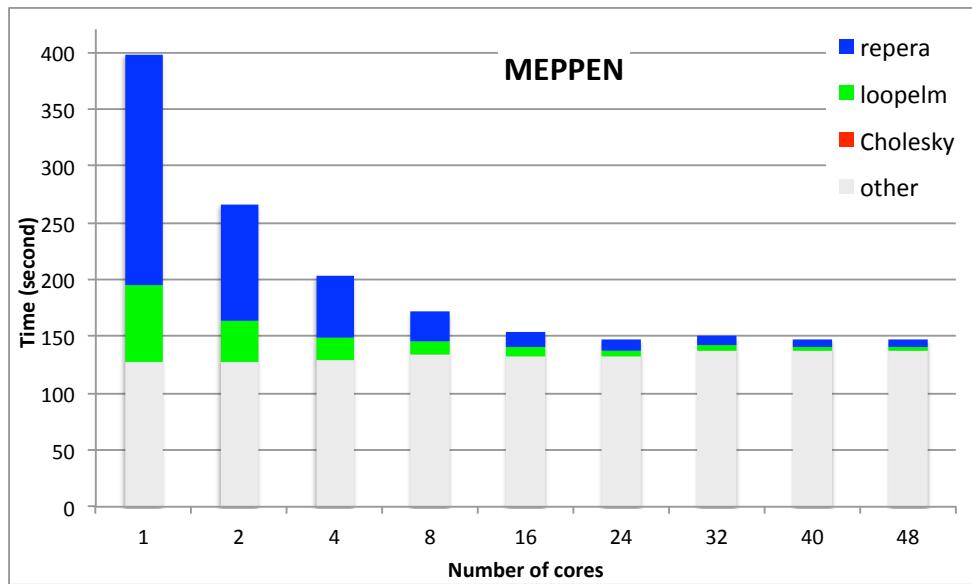


# Multicore version of EPX

- **Complex code**
  - 600 000 lines of code (Fortran + MPI)
- **Two main sources of parallelization (~70% of the computation)**
  - **Sparse Cholesky factorization**
    - skyline representation
    - dependent tasks with data flow dependencies
  - **2 Independent loops**
    - LOPELM:
      - iteration over finite elements to compute nodal internal forces
    - REPERA:
      - iteration for kinematic link detection
- **ANR REPDY**
  - **2009-2012**
  - **parallelization using XKaapi**
    - tasks with data flow dependencies
    - on demand task creation for // loop
    - Fortran API
- **Phd Student [2013-2015]**

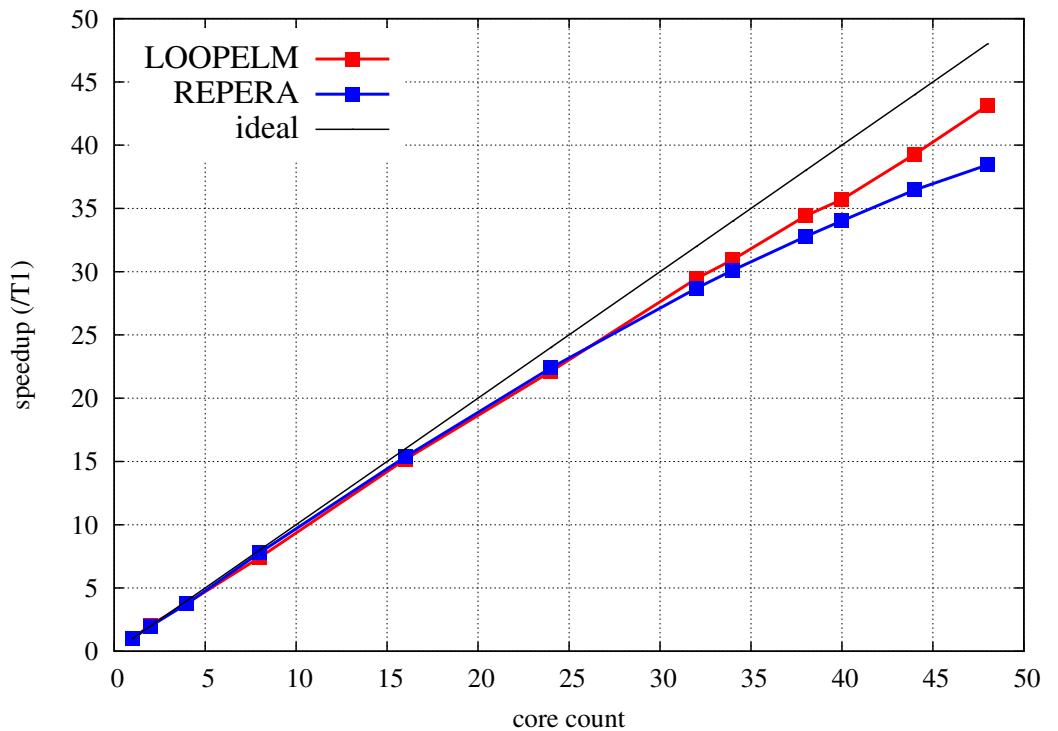
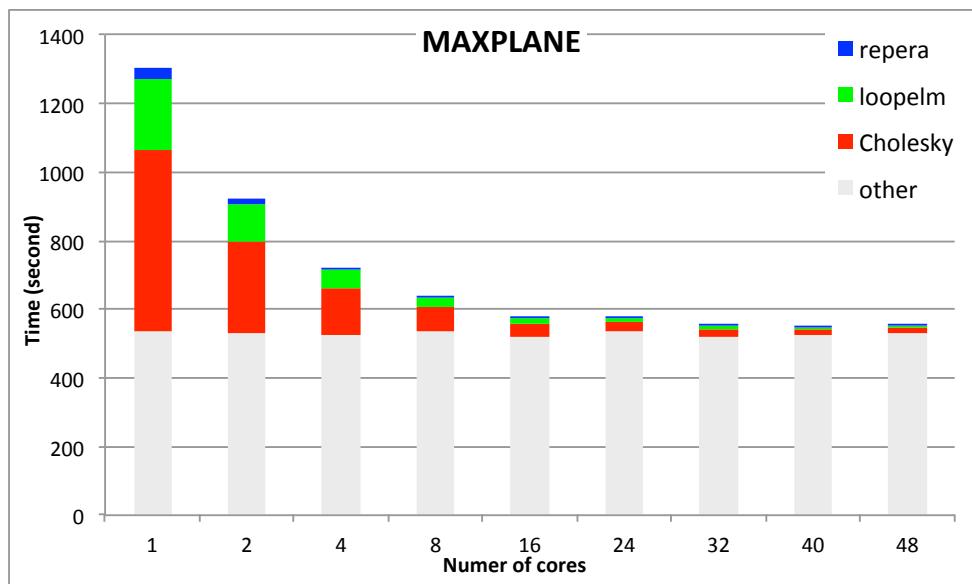
# Case of study: MEPPEN

- Main characteristics
  - Most of the time in independent loops LOOPELM and REPERA
- AMD Many Cours, 2.2GHz, 48 cores, 256GB main memory



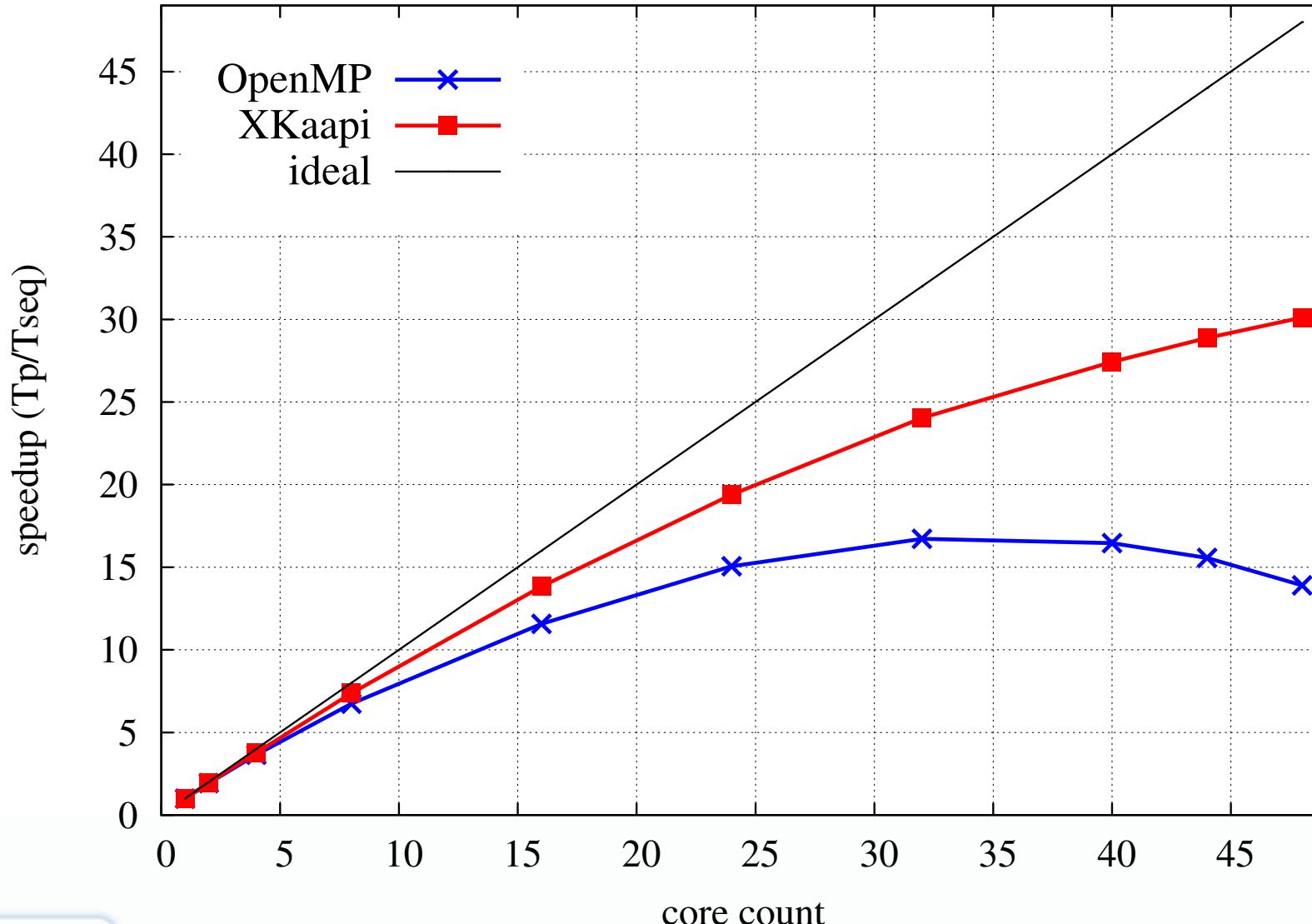
# Case of study: MAXPLANE

- Main characteristics
  - Most of the time in sparse Cholesky factorization
- AMD Many Cours, 2.2GHz, 48 cores, 256GB main memory



# Sparse Cholesky Factorization (EPX)

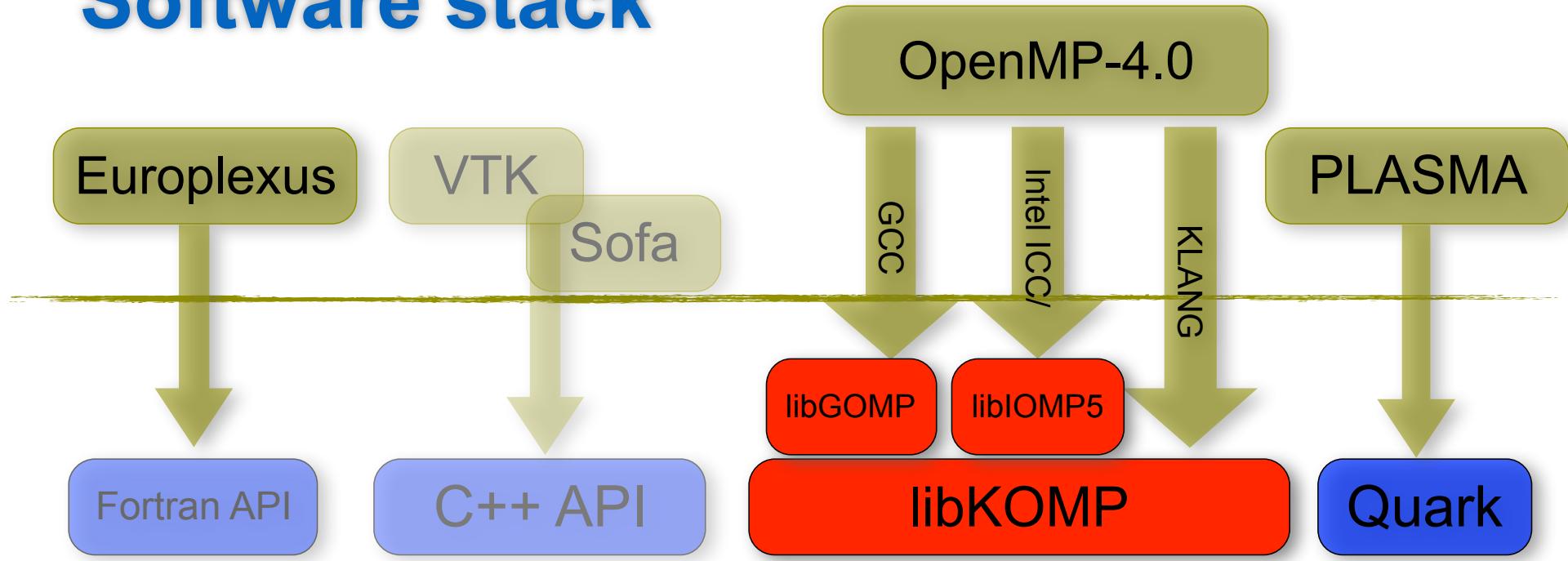
- OpenMP / XKaapi = independent tasks versus dependent tasks
  - ▶ 59462 with 3.59% of non zero elements



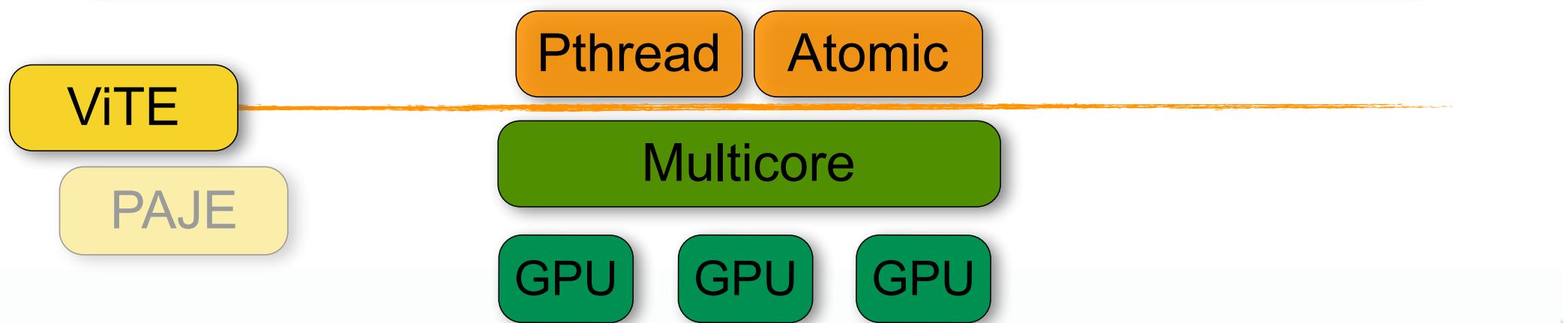
## XKaapi status

- software stack
- multi-gpu
- Intel Xeon Phi
- gdual scheduling for heterogeneous

# Software stack

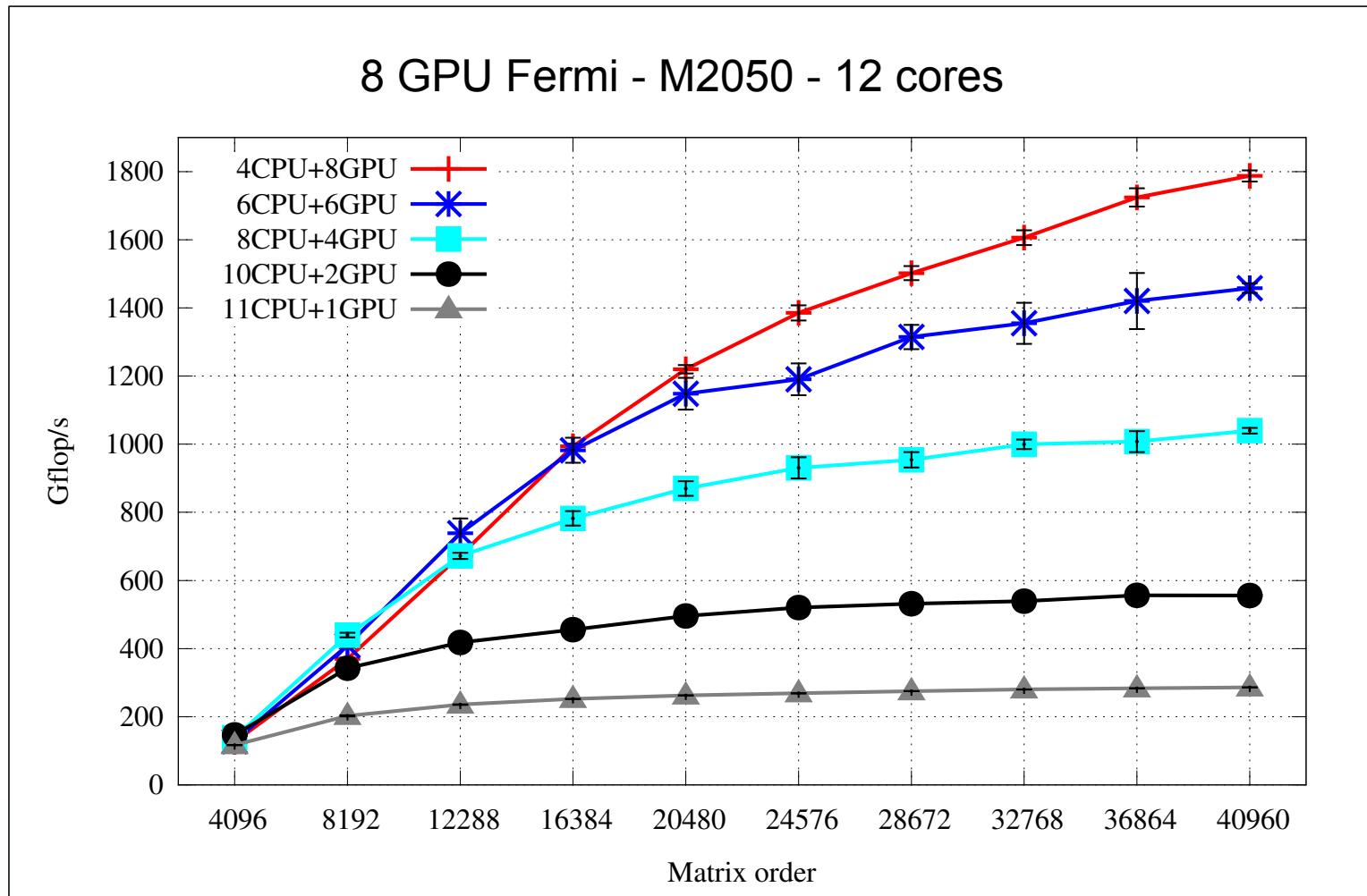


XKaapi runtime: adaptive task with data flow



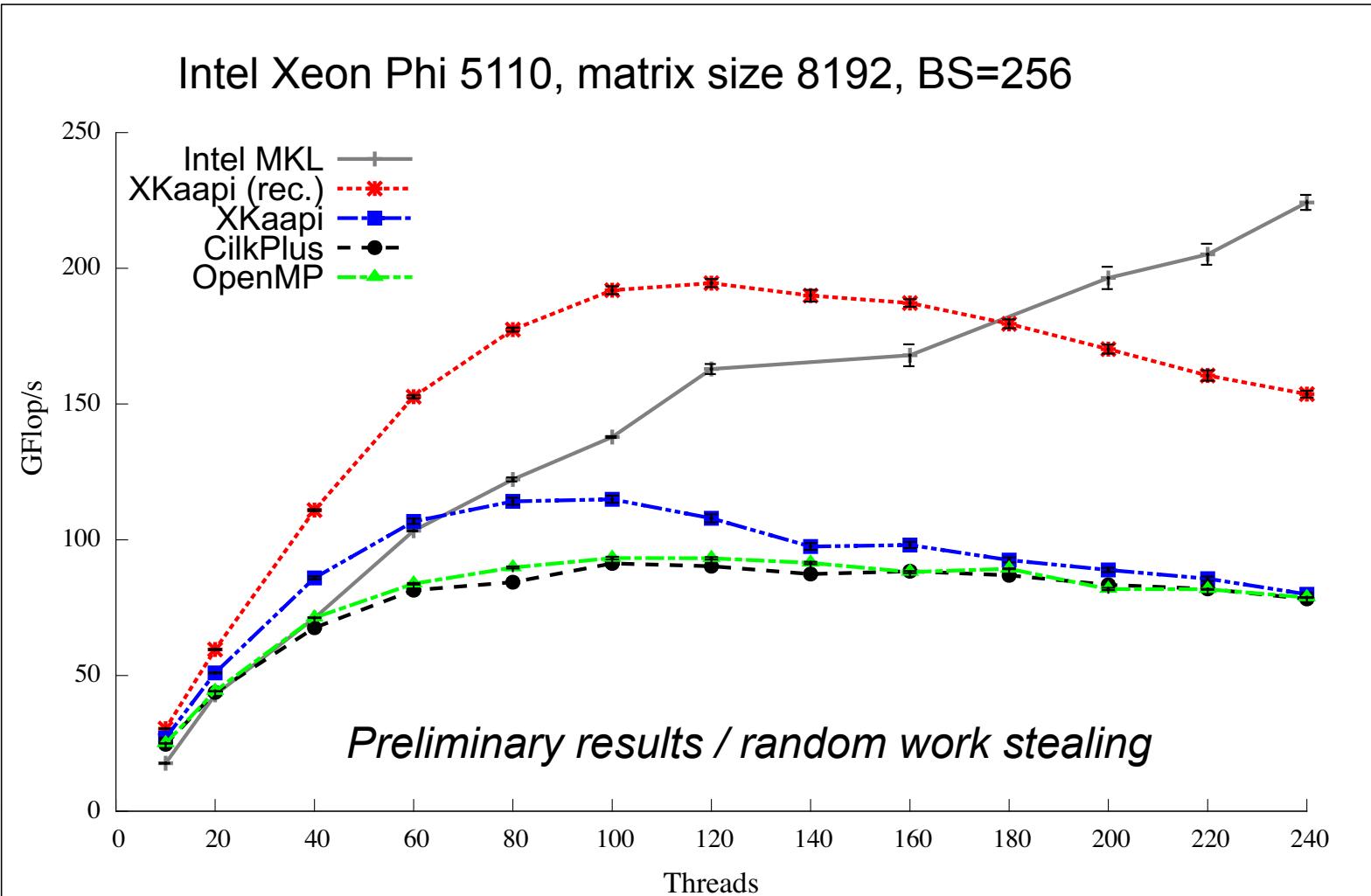
# Multi-GPUs Support

- [IPDPS 2013] work with J. Lima, N. Maillard, B. Raffin
- DPOTRF
  - ▶ using recursive tasks (on panel factorization) to reduce critical path



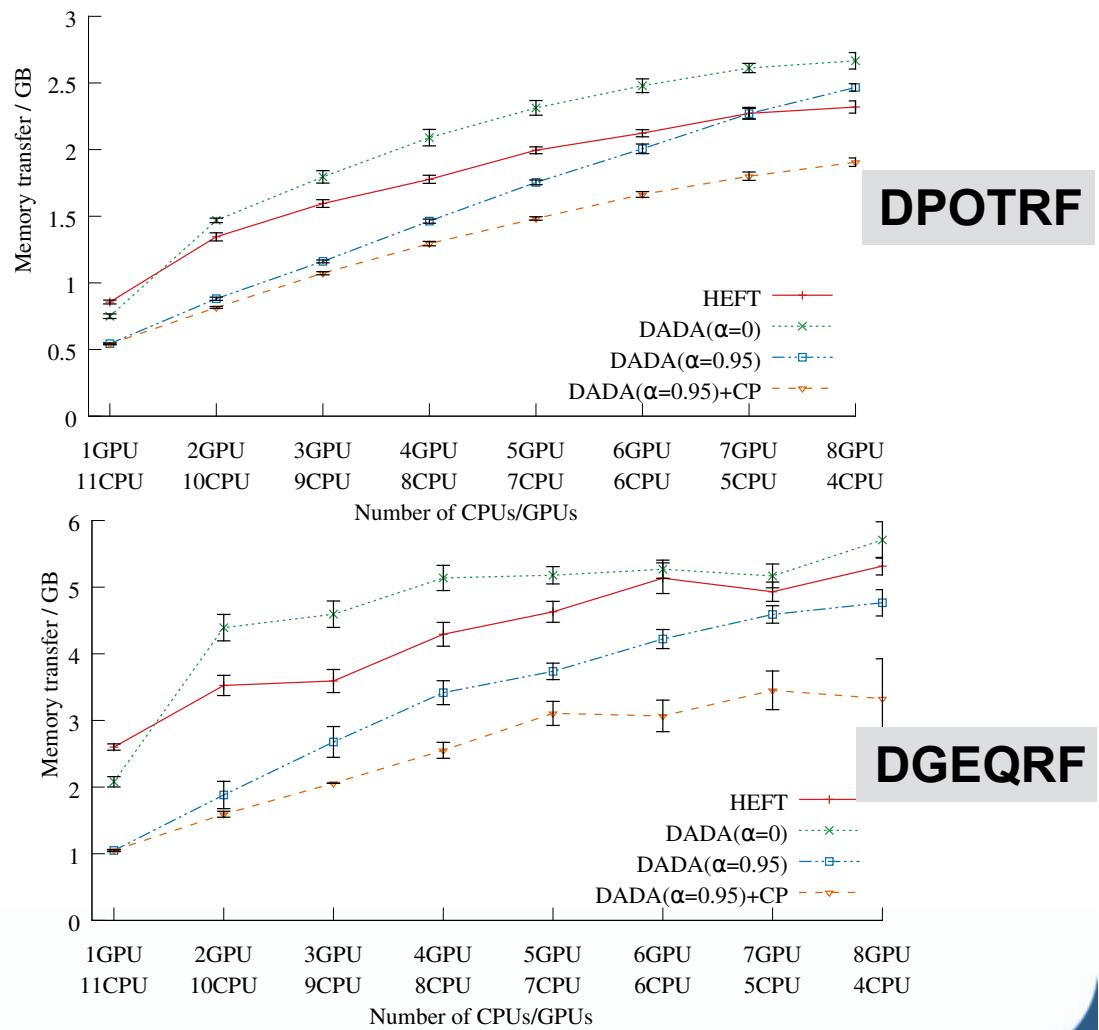
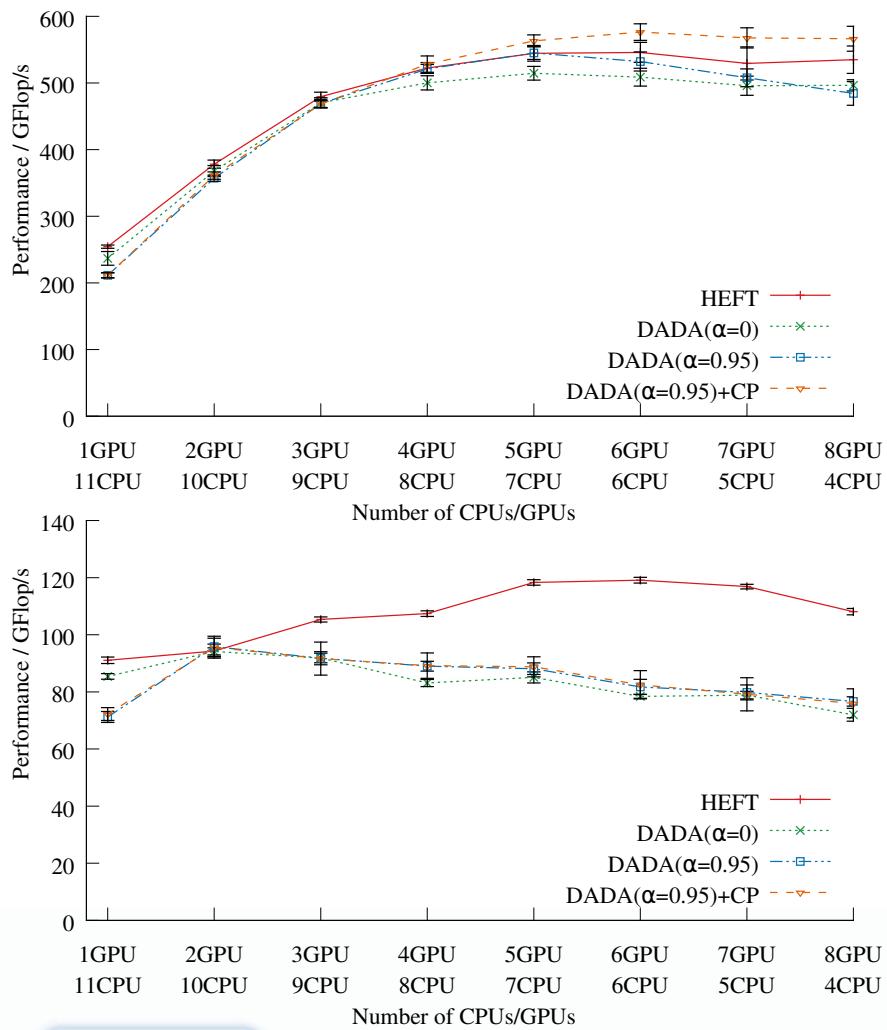
# MIC Support

- [SBAC-PAD 2013] work with F. Broquedis, J. Lima, B. Raffin
  - Experiments with XKaapi on Intel Xeon Phi Coprocessor
- DPOTRF



# Dual Approximation

- [Europar 2014], work with D. Trystram, R. Bleuse, J. Lima
  - Scheduling data flow program in XKaapi : A new affinity-based algorithm for heterogeneous architectures.
- Optimization : communication and performances



# Conclusions

- **XKaapi**

- **runtime of OpenMP + extensions**
  - libGOMP/GCC : very good support
  - libIOMP5/Intel or Clang-omp from Intel
- **adaptive task**
- **scheduling**
  - with / without « performance model »

- **Next steps**

- **programming adaptive algorithms**
  - compiler extension for OpenMP
- **improving scheduling of OpenMP task' based program**
  - temporal locality
- **scalable algorithms**
  - many-(many) core
- **reduce the size of the library**



# XKaapi History

