

Recursive Pizza



Tiled Pizza



Inner-blocking Pizza



Azzam Haidar ICL Friday talk, August 24, 2012



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General Overview: the linear algebra algorithms

Two categories:

1. One sided algorithms

- Cholesky, QR decomposition, LU factorisation.

2. Two sided algorithms

- **Eigenvalue and Singular value problems.**

Eigenvalues, eigenvectors and eigenspaces are the properties of a matrix.

- *Eigendecomposition have their origin in physics*
- Stress and strain problems
- Differential equations and quantum mechanics
- *Weather forecast*
- *Electronics simulation*
- *Image processing*
- *Material chemistry*
- *Data storage*
- *Web analysis*
- *etc...*

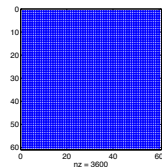
General Overview: the Eigenproblem algorithms

- Symmetric EVP $Ax = \lambda x$
 - Tri-Diagonalization Reduction + solve + back transformation.
- Generalized EVP $Ax = \lambda Bx$ or $ABx = \lambda x$
 - Cholesky + Tri-Diagonalization Reduction + solve + back transformation.
- Singular Value Decomposition $A = U\Sigma V^T$
 - Bi-Diagonalization Reduction + solve + back transformation.

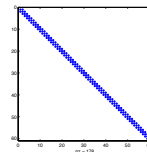
General Overview: the Eigenproblem algorithms

- Symmetric EVP $Ax = \lambda x$ meaning compute $A = Z \lambda Z^*$ where λ are the Eigenvalues and Z are the eigenvectors.

1. Tri-Diagonalization Reduction: transform A to nice form 😊



\Rightarrow



$$A = Q T Q^*$$

2. Solve: compute the Eigenvalue and Eigenvectors of the tridiagonal

$$T = E \lambda E^*$$

3. Back transformation: update the computed Eigenvectors.

$$Z = Q^* E$$

General Overview: the Eigenproblem algorithms

- 90% if only eigenvalues
- 50% if eigenvalues and eigenvectors

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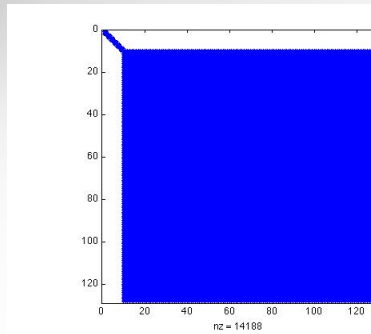
General Overview: the Eigenproblem algorithms

There are **two paths** to tridiagonal form

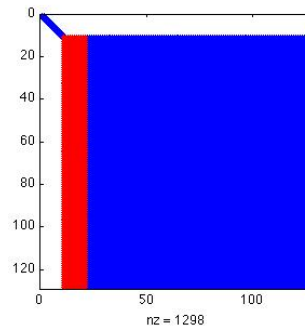
1. The standard LAPACK algorithm.
2. A new technique based on multi-stage algorithm.

Christian Bischof, Bruno Lang, Xiaobai Sun (94) proposed multiple-stage implementation called Successive Band Reductions to reduce a matrix to tridiagonal.

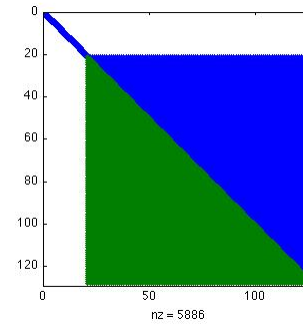
The standard Tridiagonal reduction xSYTRD



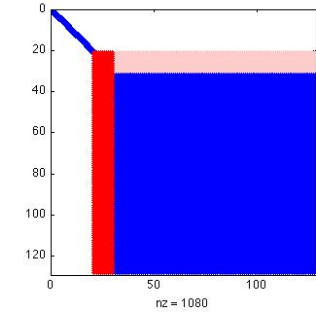
step k :



$Q \ A \ Q^H$



then update \rightarrow step $k+1$

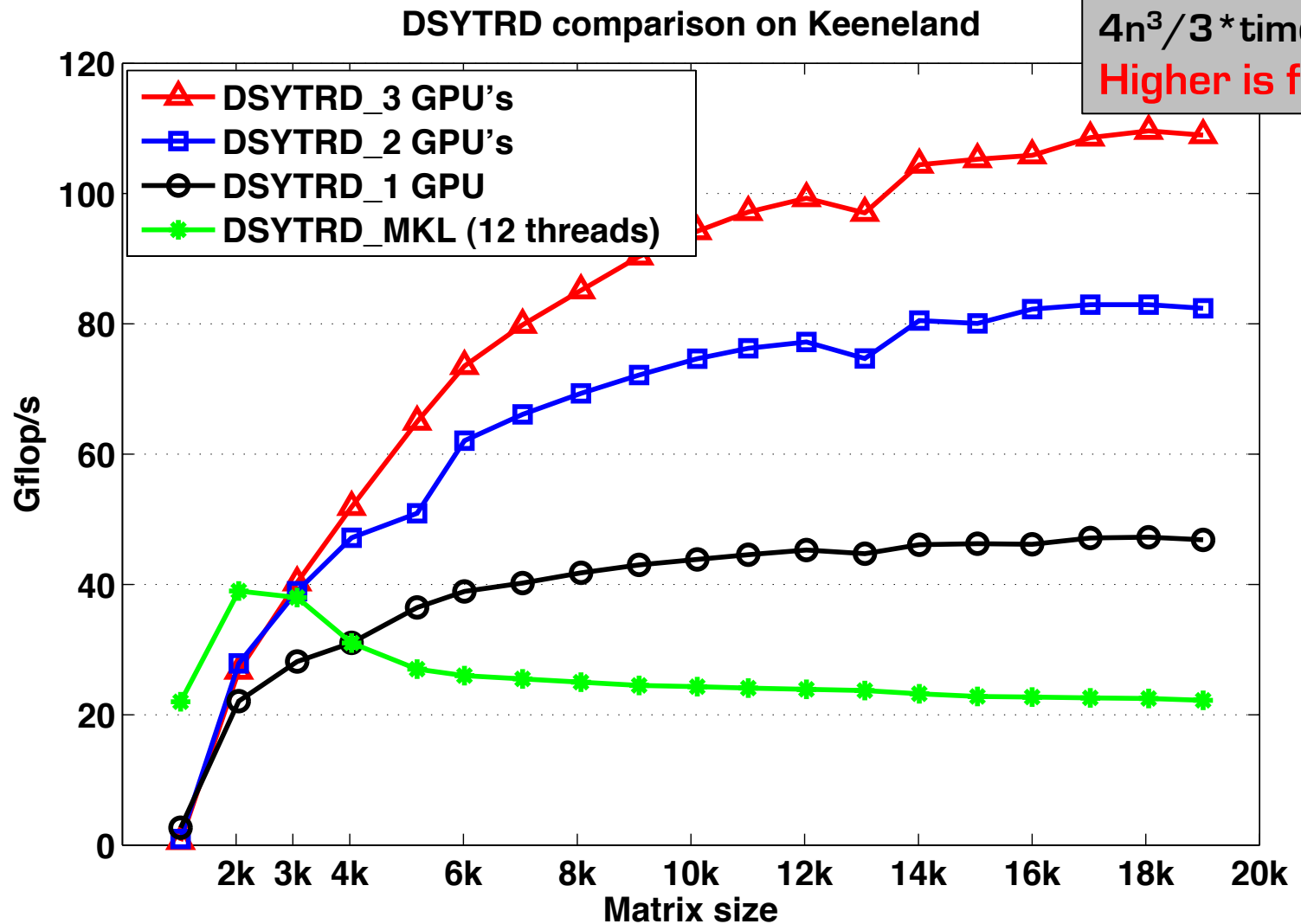


★ Characteristics

- Too many Blas-2 op,
- Relies on panel factorization,
- Total cost $4n^3/3$,
- \rightarrow Bulk sync phases,
- \rightarrow Memory bound algorithm.

The MAGMA **full** reduction to **tridiagonal**

Performance comparison



Results from Ichi

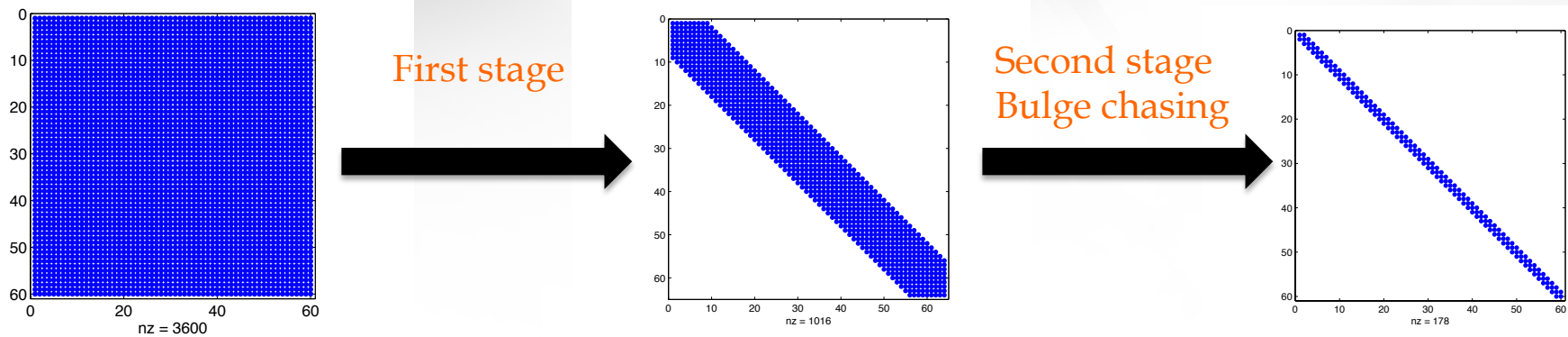
Experiments on socket of six-Intel Xeon X5650 + Fermi M2090.

The PLASMA reduction: 2 stage algorithm

Idea:

- The idea is to cast expensive memory operations, occurring during the panel factorization into fast compute intensive ones.
- Redesign the algorithm in a new fashion which increase the cache reuse.
- Design new cache friendly kernels to overcomes the memory bound limitation.
- Extract parallelism and schedule task in an asynchronous order.

The PLASMA reduction: 2 stage algorithm



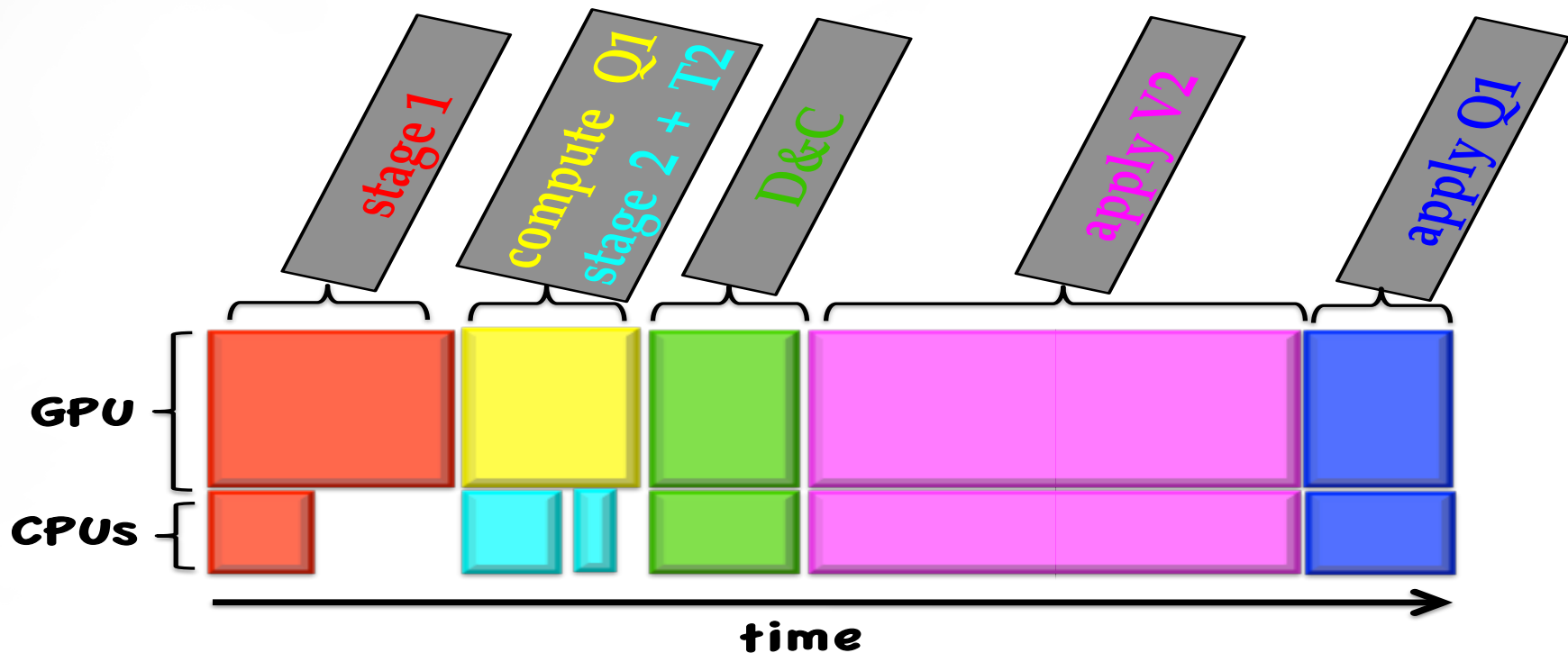
★ Characteristics

- **Stage 1:**
 - BLAS-3,
 - one shot reduction,
 - asynchronous execution,
- **Stage2:**
 - BLAS-1.5,
 - element-wise/column-wise,
 - asynchronous execution,
 - new cache friendly kernel.

MAGMA: the Eigenproblem algorithms

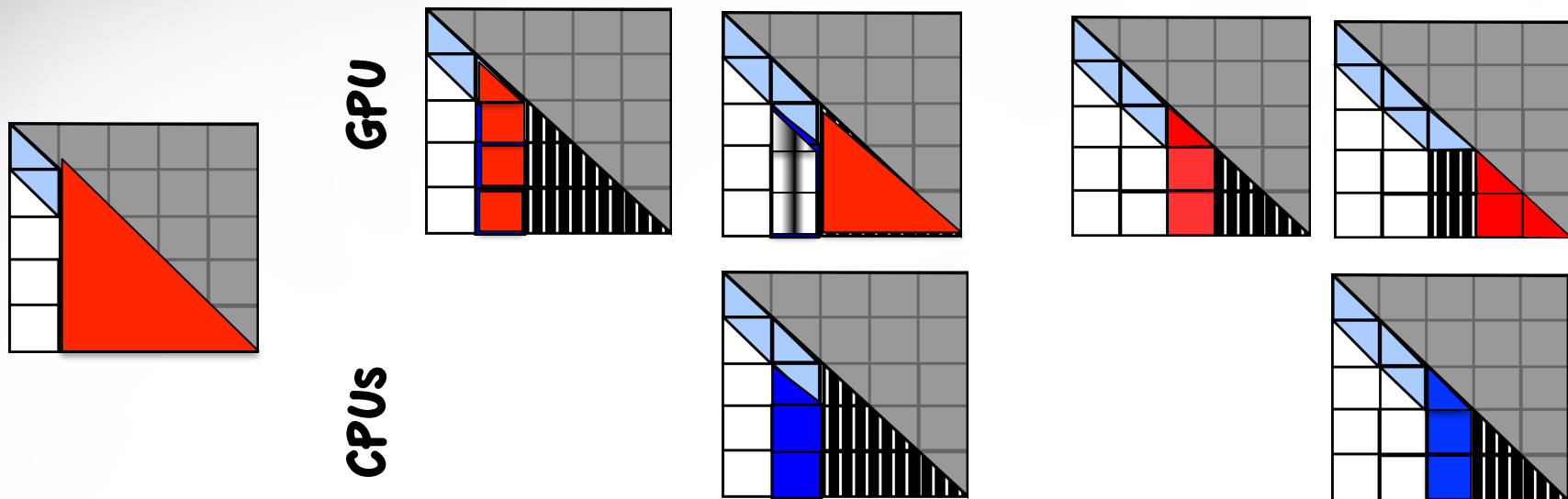
Idea:

- Develop similar approach for hybrid architectures (CPU+GPU)
- The idea is to dump expensive operations into GPU and try to overlap with the CPU.



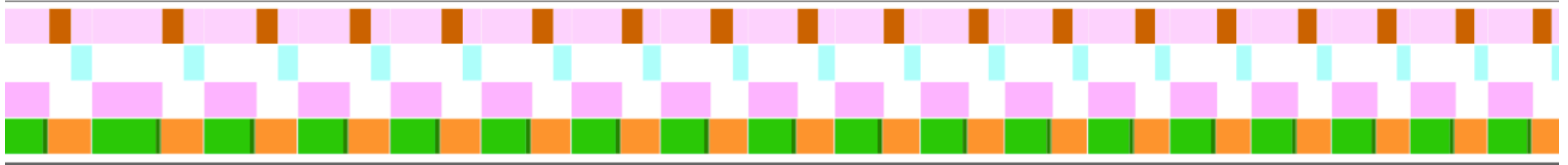
MAGMA: the Eigenproblem algorithms

The reduction:



MAGMA: the Eigenproblem algorithms

The reduction:

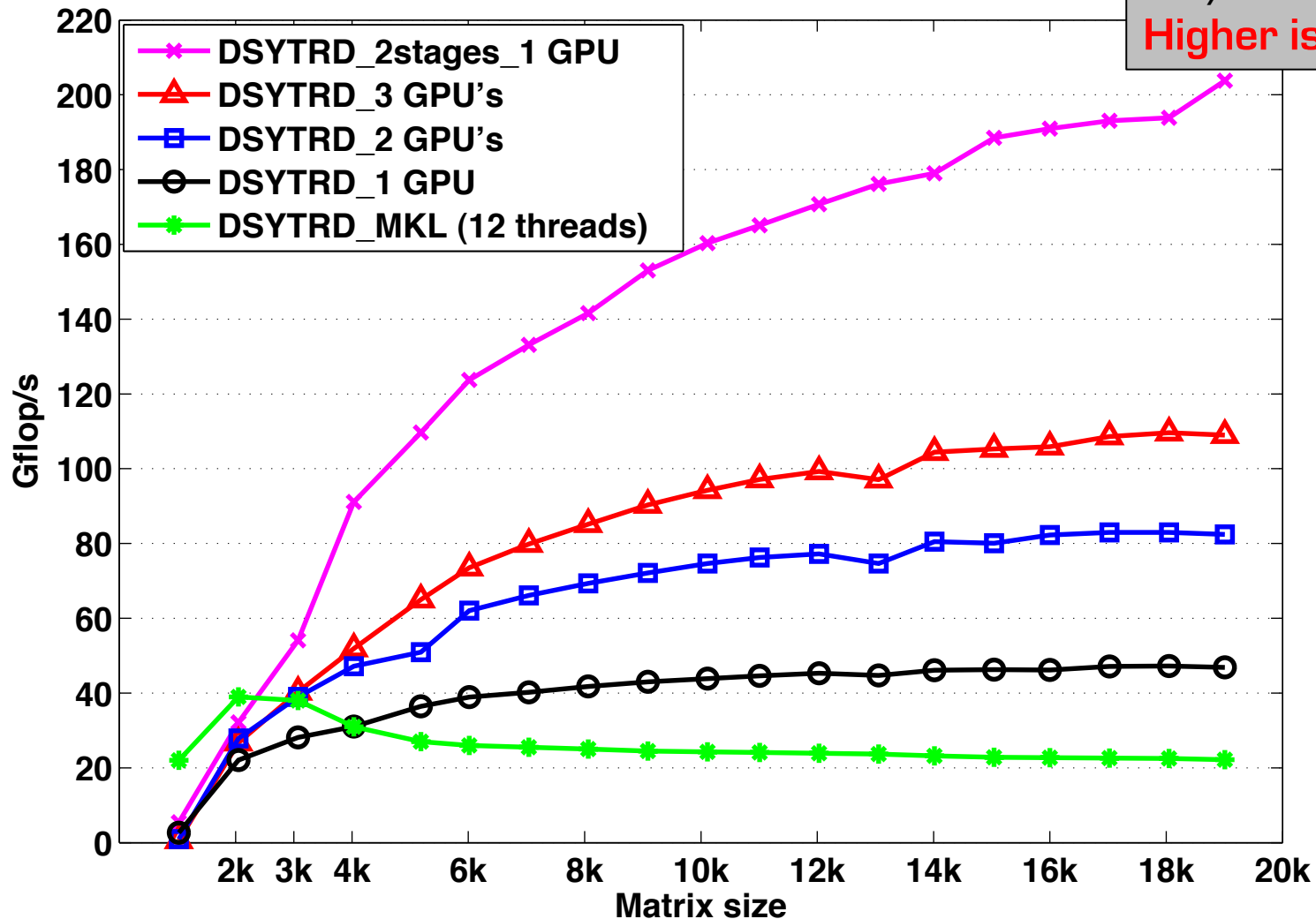


courtesy from Mark Gates

The MAGMA **full** reduction to **tridiagonal**

Performance comparison

DSYTRD comparison on Keeneland



Experiments on socket of six-Intel Xeon X5650 + Fermi M2090.

PLASMA: the Eigenproblem algorithms

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PLASMA: the Eigenproblem algorithms

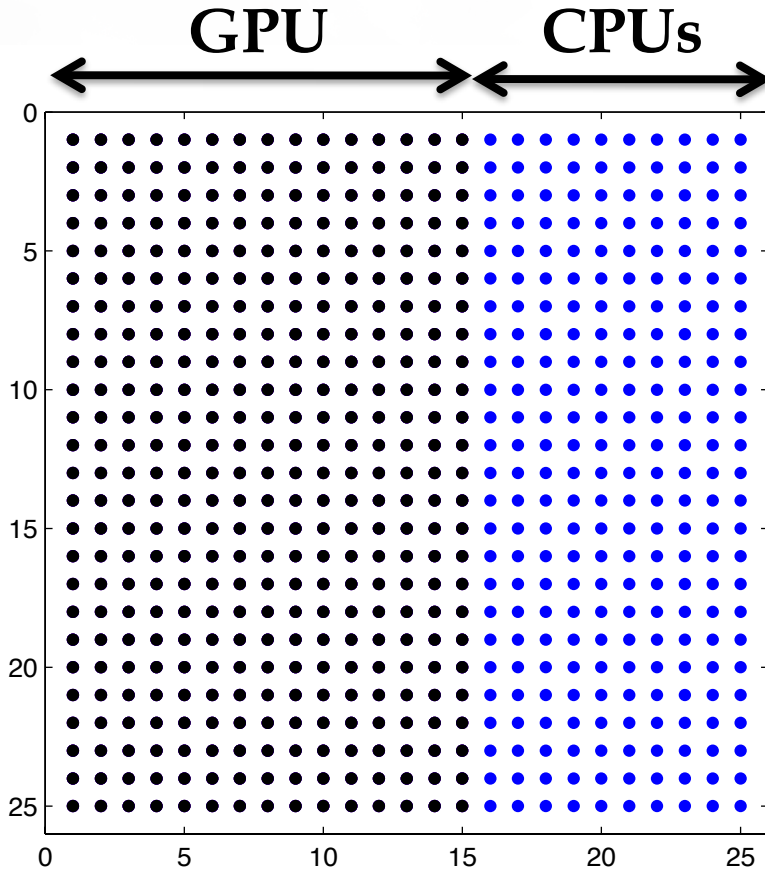
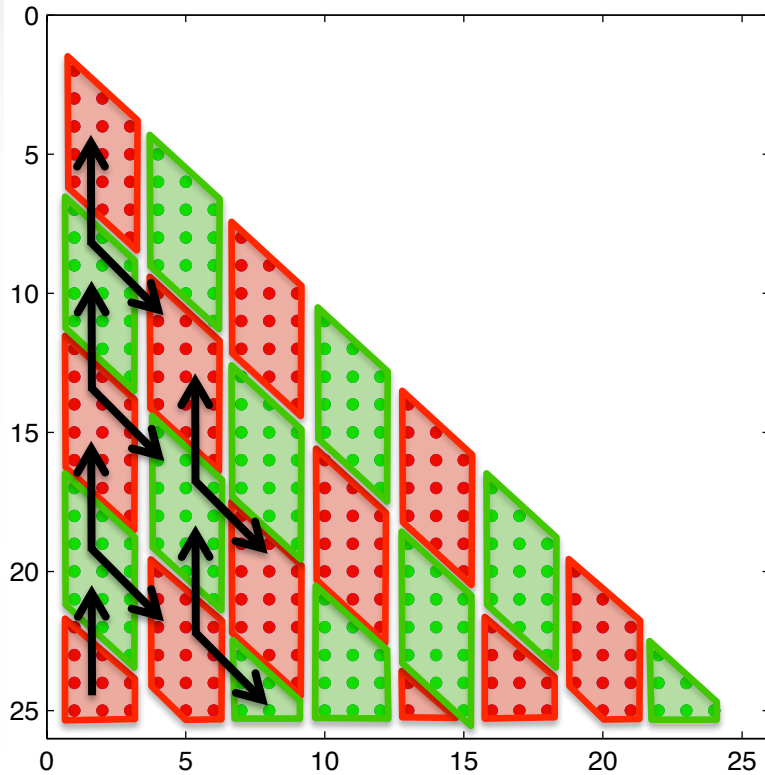
★ Characteristics

- Algorithm is more challenging, we will have 2 back transformations to apply,
- Allow independent parallelism,
- Deal with different layout of storage,
- Requires new kernels to improve cache reuse,
- is $2n^3$ more expensive.

MAGMA: the Eigenproblem algorithms

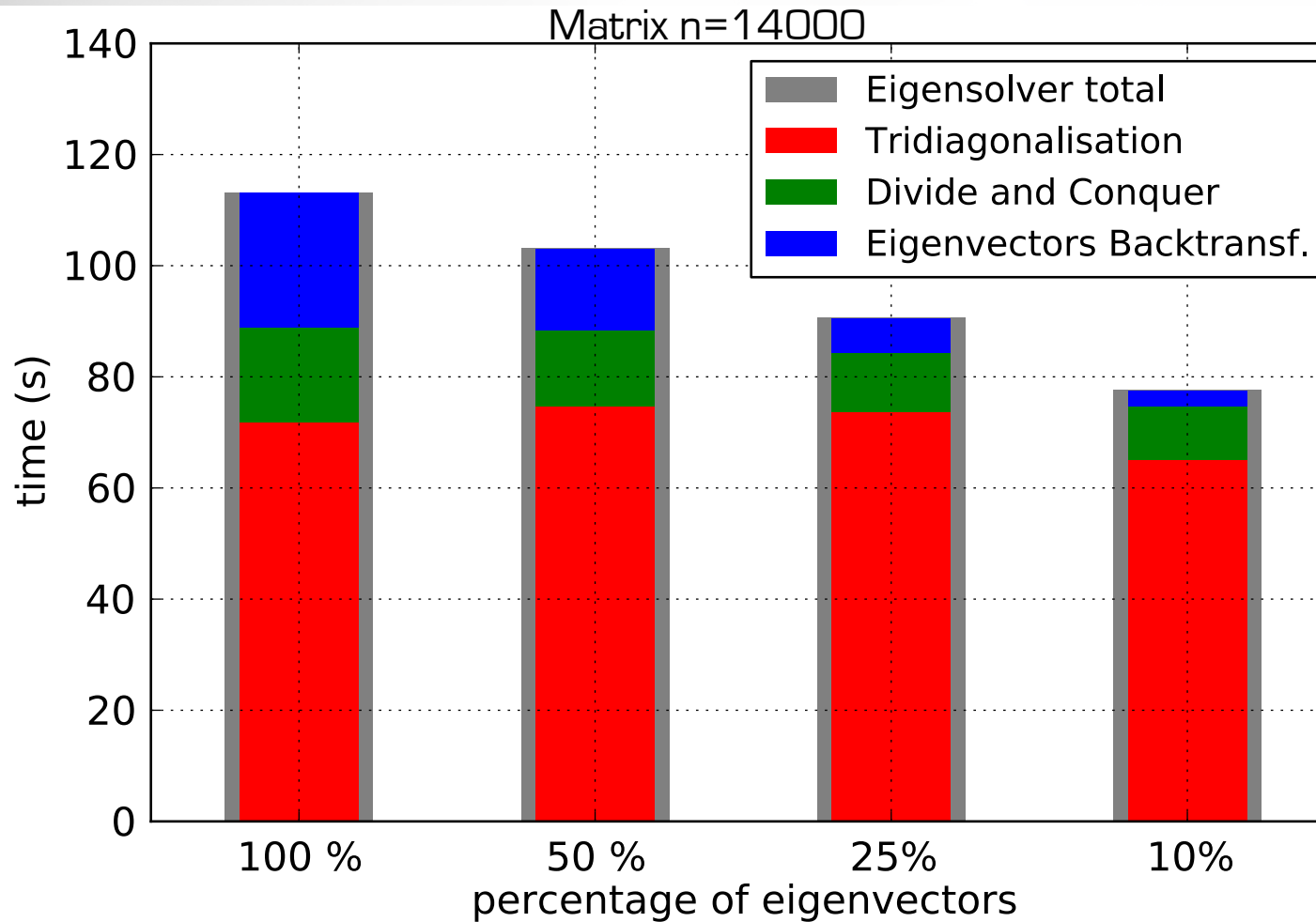
Back transformation:

- Allow CPUs to contribute



MAGMA: the Eigenproblem algorithms

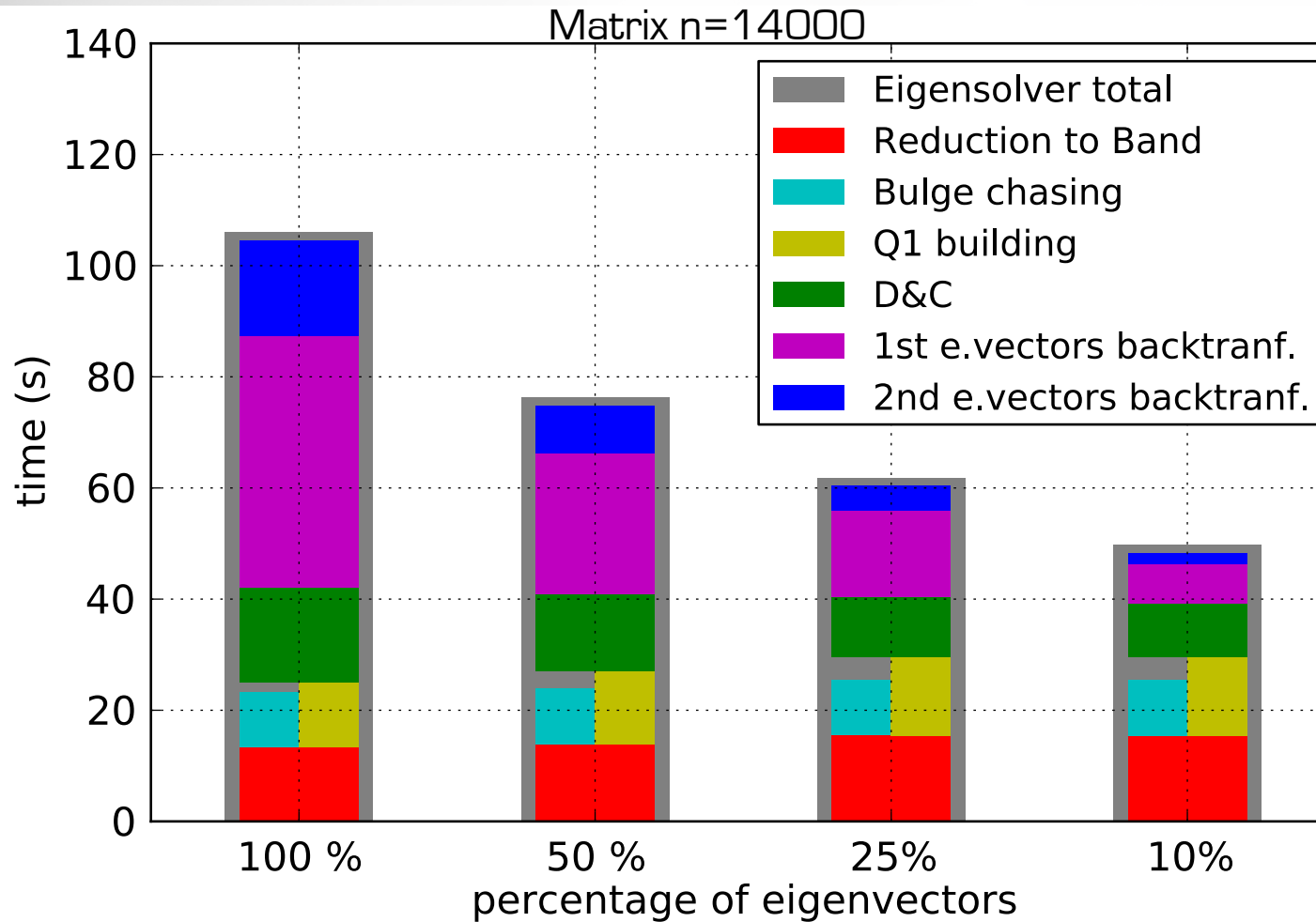
Performance comparison



Experiments on socket of six-Intel Xeon X5650 + Fermi M2090.

MAGMA: the Eigenproblem algorithms

Performance comparison

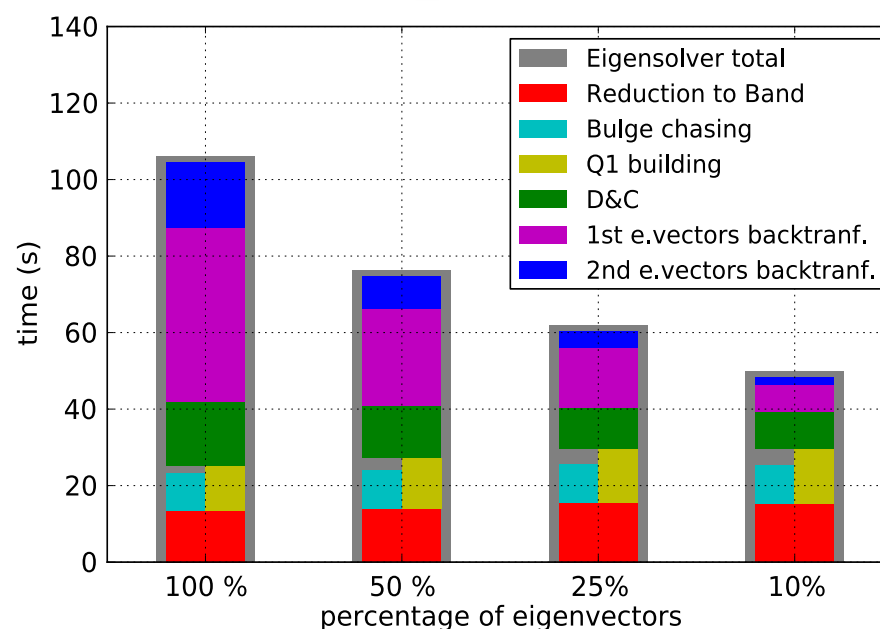
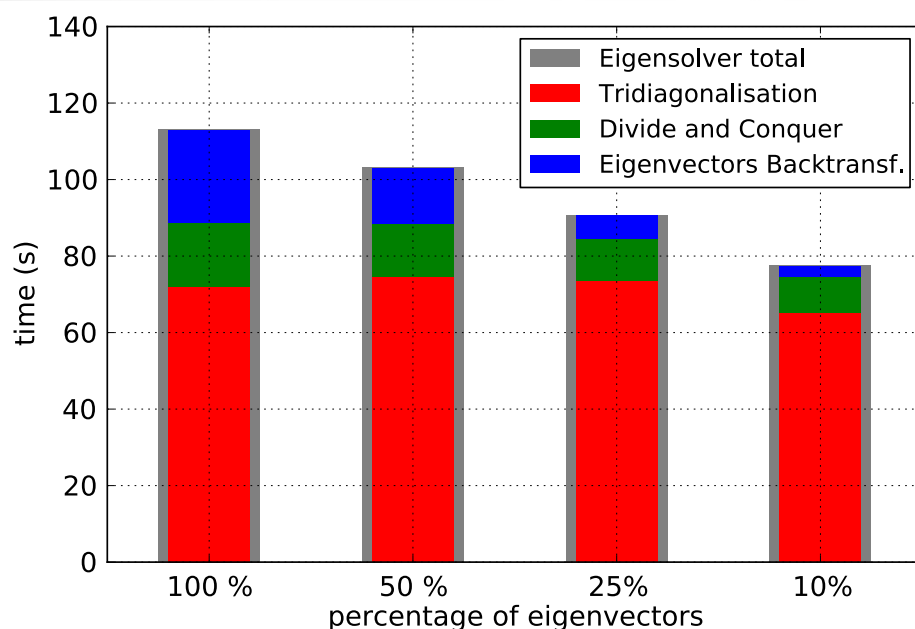


Experiments on socket of six-Intel Xeon X5650 + Fermi M2090.

MAGMA: the Eigenproblem algorithms

Performance comparison

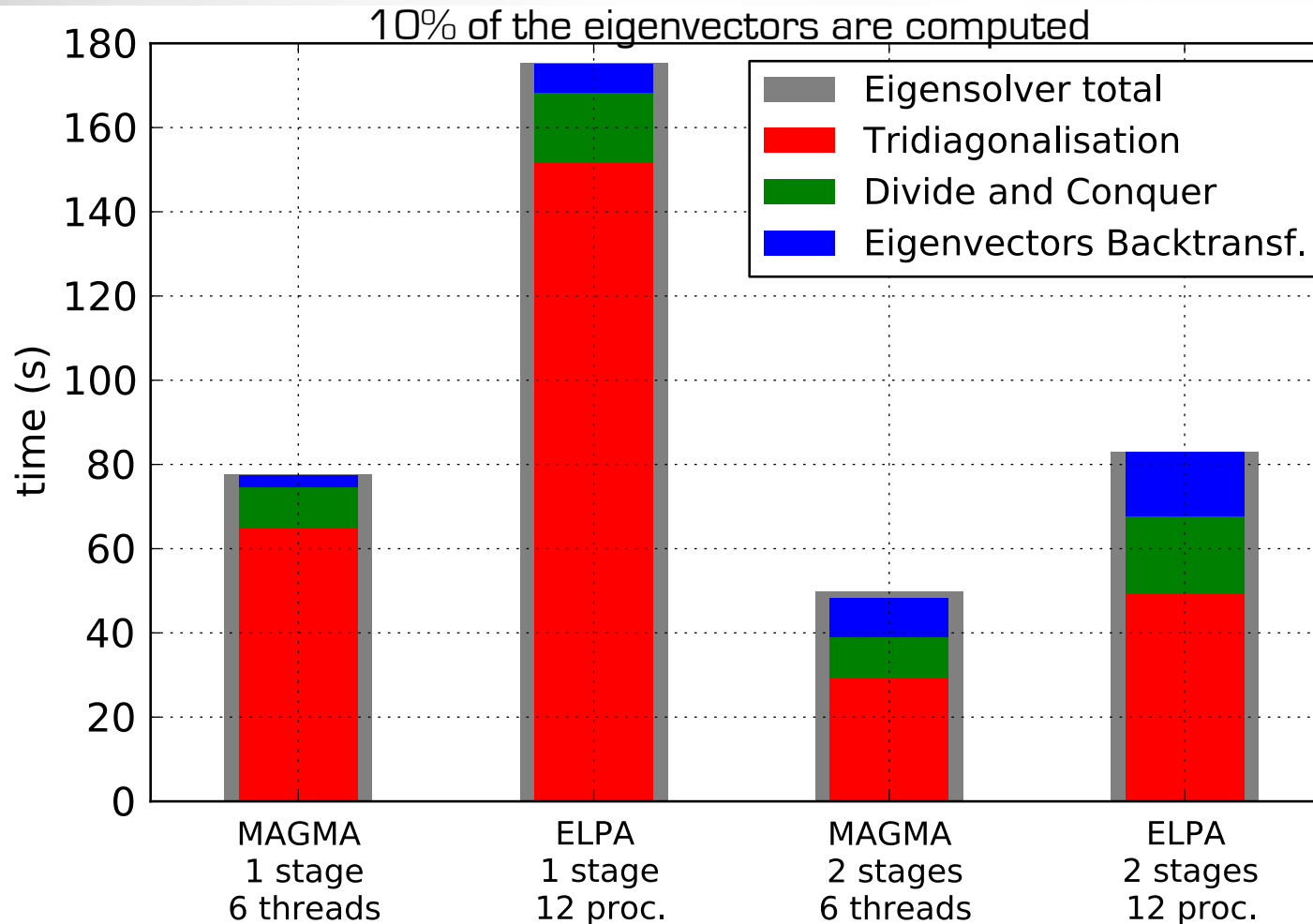
Matrix $n=14000$



Experiments on socket of six-Intel Xeon X5650 + Fermi M2090.

MAGMA: the Eigenproblem algorithms

Performance comparison



Experiments on socket of six-Intel Xeon X5650 + Fermi M2090.

Future work

★ Road map and open questions:

- Develop similar approach for SVD (ongoing integration).
- Effort might be made on the eigensolvers.
- Hessenberg, (Piotr, Hatem)
 - Bulge chasing
 - Gaussian reduction
 - Sign functions

Future work

★ Road map and open questions:

- Developing a multi-GPU/multicore/distributed version of the algorithm (ongoing).
- Develop the tridiagonal reduction in the context of distributed memory architecture.
- Evaluate a band divide and conquer for the band matrices.

Thank you for your attention