Block Locally Optimal Preconditioned Eigenvalue Xolvers

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SIAM Conference on Parallel Processing for Scientific Computing
February 24, 2006

Supported by the Lawrence Livermore National Laboratory and
the National Science Foundation
Abstract

\[ T(A - \lambda B)x = 0 \]

Block Locally Optimal Preconditioned Eigenvalue Xolvers (BLOPEX) is a package, written in C, that at present includes only one eigenxolver, Locally Optimal Block Preconditioned Conjugate Gradient Method (LOBPCG). BLOPEX supports parallel computations through an abstract layer. BLOPEX is incorporated in the HYPRE package from LLNL and is available as an external block to the PETSc package from ANL as well as a stand-alone serial library.
Acknowledgements

Supported by the Lawrence Livermore National Laboratory, Center for Applied Scientific Computing (LLNL–CASC) and the National Science Foundation.

We thank Rob Falgout, Charles Tong, Panayot Vassilevski, and other members of the Hypre Scalable Linear Solvers project team for their help and support.

We thank Jose E. Roman, a member of SLEPc team, for writing the SLEPc interface to our Hypre LOBPCG solver.

The PETSc team has been very helpful in adding our BLOPEX code as an external package to PETSc.
CONTENTS

1. Background concerning the LOBPCG algorithm
2. Hypre and PETSc software libraries
3. LOBPCG implementation strategy in BLOPEX
4. Testing
5. Scalability Results on Beowulf and BlueGene/L
6. Conclusions
LOBPCG Background
Locally Optimal Block Preconditioned Conjugate Gradient Method

The algorithm is described in:


LOBPCG Features

- LOBPCG solver finds the smallest eigenpairs of a symmetric generalized definite eigenvalue problem using preconditioning directly.

- For computing only the smallest eigenpair, the algorithm LOBPCG (Block size = 1) implements a local optimization of a 3-term recurrence.

- For finding $m$ smallest eigenpairs the Rayleigh-Ritz method on a $3m$–dimensional trial subspace is used during each iteration for the local optimization. Cluster robust, does not miss multiple eigenvalues!

- The algorithm is matrix free since the multiplication of a vector by the matrices of the eigenproblem and an application of the preconditioner to a vector are needed only as functions.
What is LOBPCG for $Ax = \lambda Bx$?

The method combines robustness and simplicity of the steepest descent method with a three-term recurrence formula:

$$x^{(i+1)} = w^{(i)} + \tau^{(i)} x^{(i)} + \gamma^{(i)} x^{(i-1)},$$

$$w^{(i)} = T(A x^{(i)} - \lambda^{(i)} B x^{(i)}),$$

$$\lambda^{(i)} = \frac{x^{(i)}, A x^{(i)}}{B x^{(i)}, x^{(i)}}$$

with properly chosen scalar iteration parameters $\tau^{(i)}$ and $\gamma^{(i)}$. The easiest and most efficient choice of parameters is based on an idea of local optimality Knyazev 1986, namely, select $\tau^{(i)}$ and $\gamma^{(i)}$ that minimize the Rayleigh quotient $\lambda(x^{(i+1)})$ by using the Rayleigh–Ritz method.

**Three-term recurrence** + **Rayleigh–Ritz method** = **Locally Optimal Conjugate Gradient Method**
Currently available LOBPCG software

- MATLAB (Knyazev)
- C in BLOPEX a standalone serial library
- C in BLOPEX built-in into Hypre and as an external package in PETSc
- SLEPc interface to Hypre LOBPCG (Jose Roman, SLEPc)
- C++ (Rich Lehoucq and Ulrich Hetmaniuk, Anasazi Trilinos)
- Fortran 77 (Randolph Bank, PLTMG)
- Python (Peter Arbenz and Roman Geus, PYFEMax)
- C++ (Sabine Zaglmayr and Joachim Schberl, NGSolve)
- Fortran 90 (Gilles Zèrah, ABINIT, complex Hermitian)
- Fortran 90 (S. Tomov and J. Langou, PESCAN, complex Hermitian)
Portable, Extensible Toolkit for Scientific Computation (PETSc) and High Performance Preconditioners (Hypre)

- Software libraries for solving large systems on massively parallel computers
- The libraries are designed to provide robustness, ease of use, flexibility and interoperability.
- The primary goal of Hypre is to provide users with advanced high-quality parallel preconditioners for linear systems.
- The primary goal of PETSc is to facilitate the integration of independently developed application modules with strict attention to component interoperability.
**BLOPEX serial/Hypre/PETSc Implementation**

- Abstract matrix- and vector-free implementation in C-language
- Hypre/PETSc and LAPACK libraries
- User-provided functions for matrix-vector multiply and preconditioner
- LOBPCG implementation utilizes Hypre/PETSc parallel vector manipulation routines
Advantages of native Hypre/PETSc implementations of LOBPCG:

- A native Hypre LOBPCG version efficiently takes advantage of powerful Hypre algebraic and geometric multigrid preconditioners.

- A native PETSc LOBPCG version gives the PETSc users community an easy access to a customizable code of the high quality modern preconditioned eigensolver and an opportunity to easily call Hypre preconditioners from PETSc.
BLOPEX Implementation Using Hypre/PETSc

- Abstract LOBPCG in C
- Interface PETSc-BLOPEX
- PETSc driver for LOBPCG
- PETSc libraries
- Interface Hypre-BLOPEX
- Hypre driver for LOBPCG
- Hypre libraries

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Domain Decomposition and Multilevel Preconditioners Tested with LOBPCG

**Hypre Implementation:**
- PFMG-PCG: geometric multigrid called directly or through PCG
- AMG-PCG: algebraic multigrid called directly or through PCG
- Schwarz-PCG: additive Schwarz called directly or through PCG

**PETSc Implementation:**
- Additive Schwarz called directly or through PCG
- Algebraic preconditioners from the Hypre package
LOBPCG Performance vs Preconditioner Iterations

7–Point 3-D Laplacian, 1,000,000 unknowns. 1 MCR node (two 2.4-GHz Pentium 4 Xeon processors and 4 GB of memory).
LOBPCG Performance vs. Block Size

Execution time as block size increases

- **PETSc version**
- **Hypre version**

7–Point 3-D Laplacian, 2,000,000 unknowns. Preconditioner: AMG.
System: Sun Fire 880, 6 CPU.
LOBPCG Scalability

Hypre, 7–Point Laplacian, 1,000,000 unknowns per node. Preconditioner: AMG. System: Beowulf (36 dual P3 1GHz 2GB nodes)
LOBPCG Scalability

7–Point Laplacian, 2,000,000 unknowns per node. Preconditioner: ASM. System: LLNL MCR, cluster of dual Pentium 4 Xeon (2.4-GHz, 4 GB) nodes.
LOBPCG Scalability

7–Point Laplacian, 2,000,000 unknowns per node. Preconditioners: AMG, PFMG. System: LLNL MCR, cluster of dual Pentium 4 Xeon (2.4-GHz, 4 GB) nodes.
### Scalability of BLOPEX-AMG on IBM BlueGene/L

<table>
<thead>
<tr>
<th>N Proc</th>
<th>N Iter.</th>
<th>Prec. setup (sec)</th>
<th>Apply Prec. (sec)</th>
<th>Lin. Alg. (sec)</th>
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<tr>
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<td>4</td>
<td>2</td>
<td>0.2</td>
</tr>
</tbody>
</table>

**Table 1: Scalability Data for 24 megapixel image segmentation**

BLOPEX in PETSc using Hypre AMG, Block size: 1, LOBPCG tolerance: $10^{-6}$. NCAR’s single-rack Blue Gene/L with 1024 compute nodes, organized in 32 I/O nodes with 32 compute nodes each. One node is a dual-core chip, containing two 700MHz PowerPC-440 CPUs and 512MB of memory. We run 1 CPU per node.
Scalability of BLOPEX-SMG on IBM BlueGene/L

<table>
<thead>
<tr>
<th>N Proc</th>
<th>Matrix Size</th>
<th>N Iter.</th>
<th>Prec. Setup (sec)</th>
<th>Solve (sec)</th>
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</thead>
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<td>10</td>
<td>7</td>
<td>74</td>
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<tr>
<td>64</td>
<td>32.768 M</td>
<td>8</td>
<td>11</td>
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<tr>
<td>512</td>
<td>0.262144 B</td>
<td>7</td>
<td>19</td>
<td>61</td>
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</table>

Table 2: Scalability for 3D Laplacian $80 \times 80 \times 80 = 512,000$ mesh per CPU

BLOPEX in Hypre struct with SMG, Block size: 1, LOBPCG tolerance: $10^{-8}$. Uniform cube partitioning. 1 CPU per node.
Scalability of BLOPEX-SMG on IBM BlueGene/L

<table>
<thead>
<tr>
<th>N Proc</th>
<th>Matrix Size</th>
<th>N Iter.</th>
<th>Prec. Setup (sec)</th>
<th>Solve (sec)</th>
</tr>
</thead>
<tbody>
<tr>
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<td>64</td>
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<td>19</td>
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<td>2360</td>
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<td>512</td>
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<td>18</td>
<td>13</td>
<td>2205</td>
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</tbody>
</table>

**Table 3:** Scalability for 3D Laplacian $50 \times 50 \times 50 = 125,000$ mesh per CPU

BLOPEX in Hypre struct with SMG, Block size: 50, LOBPCG tolerance: $10^{-4}$. Uniform cube partitioning. 1 CPU per node.
Conclusions

- BLOPEX is the only currently available package that solves eigenvalue problems using Hypre and PETSc preconditioners.

- Our abstract C implementation of the LOBPCG in BLOPEX allows easy deployment with different software packages.

- User interface routines
  - are easy to use
  - are based on Hypre/PETSc standard interfaces
  - give user an opportunity to provide matrix-vector multiply and preconditioned solver functions.

- Initial scalability results look promising.