

User's Guide for ParU, an unsymmetric multifrontal multithreaded sparse LU factorization package

Mohsen Aznaveh*, Timothy A. Davis

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Abstract

ParU is an implementation of the multifrontal sparse LU factorization method. Parallelism is exploited both in the BLAS and across different frontal matrices using OpenMP tasking, a shared-memory programming model for modern multicore architectures. The package is written in C++ and real sparse matrices are supported.

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1 Introduction

The algorithms used in ParU will be discussed in a companion paper. This document gives detailed information on the installation and use of ParU. ParU is a parallel sparse direct solver. This package uses OpenMP tasking for parallelism. ParU calls UMFPACK for the symbolic analysis phase, after that, some symbolic analysis is done by ParU itself, and then the numeric phase starts. The numeric computation is a task parallel phase using OpenMP, and each task calls parallel BLAS; i.e. nested parallelism. The performance of BLAS has a heavy impact on the performance of ParU. Moreover, the way parallel BLAS can be called in a nested environment can also be very important for ParU's performance.

1.0.1 Instructions on using METIS

SuiteSparse uses METIS 5.1.0, distributed along with SuiteSparse itself. Its use is optional, however. ParU is using AMD as the default ordering. METIS tends to give orderings that are good for parallelism. However, the METIS itself can be slower than AMD. As a result, the symbolic analysis using METIS can be slow, but usually, the factorization is faster. Therefore, depending on your use case, either use METIS, or you can compile and run your code without using METIS. If you are using METIS on an unsymmetric case, UMFPACK has to form the Matrix $A^T A$. This matrix can be a dense matrix and takes a lot of resources to form it. To avoid such conditions, you can use the ordering strategy `UMFPACK_ORDERING_METIS_GUARD` that is introduced in UMFPACK version 6.0. This ordering strategy use COLAMD instead of METIS in those cases.

*email: aznaveh@tamu.edu. <http://www.suitesparse.com>.

2 Using ParU in C and C++

ParU relies on CHOLMOD for its basic sparse matrix data structure, a compressed sparse column format. CHOLMOD provides interfaces to the AMD, COLAMD, and METIS ordering methods and many other functions. ParU also relies on UMFPACK Version 6.0 or higher for symbolic analysis.

2.1 Installing the C/C++ library on any system

All of SuiteSparse can be built by `cmake` with a single top-level `CMakeLists.txt` file. In addition, each package (including ParU) has its own `CMakeLists.txt` file to build that package. This is the simplest method for building ParU and its dependent packages on all systems.

2.2 Installing the C/C++ library on Linux/Unix

In Linux/MacOs, type `make` at the command line in either the `SuiteSparse` directory (which compiles all of SuiteSparse) or in the `SuiteSparse/ParU` directory (which just compiles ParU and the libraries it requires). ParU will be compiled; you can type `make demos` to run a set of simple demos.

The use of `make` is optional. The top-level `ParU/Makefile` is a simple wrapper that uses `cmake` to do the actual build.

To fully test the coverage of the lines ParU, go to the `Tcov` directory and type `make`. This will work for Linux only.

To install the shared library (by default, into `/usr/local/lib` and `/usr/local/include`), do `make install`. To uninstall, do `make uninstall`. For more options, see the `ParU/README.md` file.

2.3 C/C++ Example

The C++ interface is written using only real matrices (double precision).

The simplest function computes the MATLAB equivalent of $\mathbf{x}=\mathbf{A}\backslash\mathbf{b}$ and is almost as simple as MATLAB.

Below is a simple C++ program that illustrates the use of ParU. The program reads in a problem from `stdin` in MatrixMarket format [3], solves it, and prints the norm of \mathbf{A} and the residual. Some error testing code is omitted to simplify showing how the program works. The full program can be found in `ParU/Demo/paru_simple.cpp`

```
#include <iostream>
#include <iomanip>
#include <iostream>
#include <cmath>
#include "ParU.h"

int main(int argc, char **argv)
```

```

{
    cholmod_common Common, *cc;
    cholmod_sparse *A;
    ParU_Symbolic *Sym;
    //~~~~~Reading the input matrix and test if the format is OK~~~~~
    // start CHOLMOD
    cc = &Common;
    int mtype;
    cholmod_l_start(cc);
    A = (cholmod_sparse *)cholmod_l_read_matrix(stdin, 1, &mtype, cc);
    //~~~~~Starting computation~~~~~
    std::cout << "===== ParU, a simple demo: =====\n";
    ParU_Control Control;
    ParU_Analyze(A, &Sym, &Control);
    std::cout << "Input matrix is " << Sym->m << "x" << Sym->n
        << " nnz = " << Sym->anz << std::endl;
    ParU_Numeric *Num;
    ParU_Factorize(A, Sym, &Num, &Control);

    //~~~~~Computing the residual, norm(b-Ax) ~~~~~
    int64_t m = Sym->m;
    double *b = (double *)malloc(m * sizeof(double));
    double *xx = (double *)malloc(m * sizeof(double));
    for (int64_t i = 0; i < m; ++i) b[i] = i + 1;
    ParU_Solve(Sym, Num, b, xx, &Control);
    double resid, anorm, xnorm;
    ParU_Residual(A, xx, b, resid, anorm, xnorm, &Control);
    double rresid = (anorm == 0 || xnorm == 0) ? 0 : (resid/(anorm*xnorm));
    std::cout << std::scientific << std::setprecision(2)
        << "Relative residual is |" << rresid << "| anorm is " << anorm
        << ", xnorm is " << xnorm << " and rcond is " << Num->rcond << "."
        << std::endl;
    free(b);
    free(xx);

    //~~~~~End computation~~~~~
    ParU_FreeNumeric(&Num, &Control);
    ParU_FreeSymbolic(&Sym, &Control);
    cholmod_l_free_sparse(&A, cc);
    cholmod_l_finish(cc);
}

```

A simple demo for the C interface is shown next. You can see the complete demo in `ParU/Demo/paru_simplec.c`

```
#include <stdint.h>
```

```

#include <math.h>
#include "ParU.h"

int main(int argc, char **argv)
{
    cholmod_common Common, *cc;
    cholmod_sparse *A;
    ParU_C_Symbolic *Sym;
    //~~~~~Reading the input matrix and test if the format is OK~~~~~
    // start CHOLMOD
    cc = &Common;
    int mtype;
    cholmod_l_start(cc);
    // A = mread (stdin) ; read in the sparse matrix A
    A = (cholmod_sparse *)cholmod_l_read_matrix(stdin, 1, &mtype, cc);
    //~~~~~Starting computation~~~~~
    printf("===== ParU, a simple demo, using C interface : =====\n");
    ParU_C_Control Control;
    ParU_C_Init_Control(&Control);
    ParU_C_Analyze(A, &Sym, &Control);
    printf("Input matrix is %" PRId64 "x%" PRId64 " nnz = %" PRId64 " \n",
        Sym->m, Sym->n, Sym->anz);
    ParU_C_Numeric *Num;
    ParU_C_Factorize(A, Sym, &Num, &Control);

    //~~~~~Computing Ax = b ~~~~~
    int64_t m = Sym->m;
    double *b = (double *)malloc(m * sizeof(double));
    double *xx = (double *)malloc(m * sizeof(double));
    for (int64_t i = 0; i < m; ++i) b[i] = i + 1;
    ParU_C_Solve_Axb(Sym, Num, b, xx, &Control);
    double resid, anorm, xnorm;
    ParU_C_Residual_bAx(A, xx, b, &resid, &anorm, &xnorm, &Control);
    double rresid = (anorm == 0 || xnorm == 0) ? 0 : (resid/(anorm*xnorm));
    printf("Relative residual is |%.2e|, anorm is %.2e, xnorm is %.2e, "
        " and rcond is %.2e.\n",
        rresid, anorm, xnorm, Num->rcond);
    free(b);
    free(xx);

    //~~~~~End computation~~~~~
    ParU_C_FreeNumeric(&Num, &Control);
    ParU_C_FreeSymbolic(&Sym, &Control);
    cholmod_l_free_sparse(&A, cc);
    cholmod_l_finish(cc);
}

```

}

2.4 C/C++ Syntax

All ParU C and C++ routines return an enum of type `ParU_Info`. The user should check the output before continuing and computing further the result of prior routine. You can see the user callable routines in `ParU/Include/ParU.h`. The following is a list of user-callable C++ functions, their prototypes, and a description of what they do:

FIXME: add prototypes and description of each parameter

1. `ParU_Version`: return the version of the ParU package you are using.
2. `ParU_Analyze`: Symbolic analysis is done in this routine. UMFPACK is called here and after that, some more specialized symbolic computation is done for ParU. `ParU_Analyze` is called once and can be used for different `ParU_Factorize` calls for the matrices that have the same pattern.
3. `ParU_Factorize`: Numeric factorization is done in this routine. Scaling and making Sx (scaled and staircase structure) matrix, computing factors, and permutations are here. `ParU_Symbolic` structure which is computed in `ParU_Analyze` is an input in this routine.
4. `ParU_Solve`: Using symbolic analysis and factorization phase output to solve $Ax = b$. In all the solve routines Num structure must come with the same Sym struct that comes from `ParU_Factorize`. This routine is overloaded and can solve different systems. It has versions that keep a copy of x or overwrite it. Also, it can solve multiple right-hand side problems.
5. `ParU_LSolve`:
6. `ParU_USolve`:
7. `ParU_Perm`:
8. `ParU_InvPerm`:
9. `ParU_Residual`: This function computes $|Ax - b|$ one norm of matrix A and one norm of x (or X for multiple right handside).
10. `ParU_FreeNumeric`: frees the numerical part of factorization.
11. `ParU_FreeSymbolic`: frees the symbolic part of factorization.

The C interface is quite similar to the C++ interface, and you can see the C user callable routines in `ParU/Include/ParU.h`. The following is a list of user-callable C functions, their prototypes, and what they can do:

1. `ParU_C_Version`: return the version of the ParU package you are using.

2. `ParU_C_Init_Control`: Initialize C Control object before using it.
3. `ParU_C_Analyze`: Symbolic analysis is done in this routine. UMFPACK is called here; after that, some more specialized symbolic computation is done for ParU. `ParU_C_Analyze` is called once and can be used for different `ParU_C_Factorize` calls for the matrices that have the same pattern.
4. `ParU_C_Factorize`: Numeric factorization is done in this routine. Scaling and making Sx (scaled and staircase structure) matrix, computing factors, and permutations are here. `ParU_C_Symbolic` structure which is computed in `ParU_C_Analyze` is an input in this routine.
5. `ParU_C_Solve_Axx`, `ParU_C_Solve_Axb`, `ParU_C_Solve_XXX` and `ParU_C_Solve_AXB`, Using symbolic analysis and factorization phase output to solve $Ax = b$. In all the solve routines Num structure must come with the same Sym struct that comes from `ParU_C_Factorize`.
6. `ParU_C_LSolve`:
7. `ParU_C_USolve`:
8. `ParU_C_Perm`:
9. `ParU_C_InvPerm`:
10. `ParU_C_Residual_bAx`: This function computes $|Ax - b|$ one norm of matrix A and one norm of x
11. `ParU_C_Residual_BAX`: This function computes $|AX - B|$ one norm of matrix A and one norm of X
12. `ParU_C_FreeNumeric`: frees the numerical part of factorization.
13. `ParU_C_FreeSymbolic`: frees the symbolic part of factorization.

2.5 Details of the C/C++ Syntax

For further details on how to use the C/C++ syntax, please refer to the definitions and descriptions in the `SuiteSparse/ParU/Include/ParU.h` file, which describes each C++ and C function. Only `double` and square matrices are supported.

There are C/C++ options to control ParU, which is an input argument to several routines. When you make C++ `ParU_Control` object, it is initialized with default values. The user can change the values. When using C `ParU_C_Control`, you have to fully initialize it or call `ParU_C_Init_Control` before using it.

Here is the list of control options (both in C and C++):

ParU_Control	default value	explanation
mem_chunk	1024 * 1024	chunk size for memset and memcpy
paru_max_threads	0	initialized with <code>omp_max_threads</code>
umfpack_ordering	UMFPACK_ORDERING_AMD	default UMFPACK ordering
umfpack_strategy	UMFPACK_STRATEGY_AUTO	default UMFPACK strategy
umfpack_default_singleton	1	default filter singletons if true
relaxed_amalgamation_threshold	32	threshold for relaxed amalgamation
scale	1	if 1 matrix will be scaled using <code>max_row</code>
panel_width	32	width of panel for dense factorization
paru_strategy	PARU_STRATEGY_AUTO	default strategy for ParU
piv_tol	0.1	tolerance for accepting sparse pivots
diag_tol	0.001	tolerance for accepting symmetric pivots
trivial	4	Do not call BLAS for smaller dgemms
worthwhile_dgemm	512	dgemms bigger than worthwhile are tasked
worthwhile_trsm	4096	trsm bigger than worthwhile are tasked

The first row of the options is either used in symbolic or numerical analysis. The second row of the options is used in the symbolic analysis. In the symbolic analysis phase, only the matrix pattern is probed. The third row of control options shows those that have an impact on numerical analysis.

`paru_max_threads` is initialized by `omp_max_threads` if the user do not provide a smaller number.

If `paru_strategy` is set to `PARU_STRATEGY_AUTO` ParU uses the same strategy as UMFPACK. However, the user can ask UMFPACK for an unsymmetric strategy but use a symmetric strategy for ParU. Usually, UMFPACK chooses a good ordering; however, there might be cases where users prefer unsymmetric ordering on UMFPACK but symmetric computation on ParU.

3 Requirements and Availability

ParU requires several Collected Algorithms of the ACM: CHOLMOD [4, 7] (version 1.7 or later), AMD [1, 2], COLAMD [5, 6] and UMFPACK [8] for its ordering/analysis phase and for its basic sparse matrix data structure, and the BLAS [9] for dense matrix computations on its frontal matrices. An efficient implementation of the BLAS is strongly recommended, either vendor-provided (such as the Intel MKL, the AMD ACML, or the Sun Performance Library) or other high-performance BLAS such as those of [10]. Note that while ParU uses nested parallelism heavily the right options for the BLAS library must be chosen to get a good performance.

The use of OpenMP tasking is optional, but without it, only parallelism within the BLAS can be exploited (if available). See ParU/Doc/LICENSE for the license. Alternative licenses are also available; contact the authors for details.

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