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

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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.

#### 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. However, depending on the input problem performance of parallelism in BLAS sometimes can have less effects in ParU.

#### 1.0.1 Instructions on using METIS

SuiteSparse is now on 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 slow. 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.

Note that METIS is not bug-free; it can occasionally cause segmentation faults, particularly if used when finding basic solutions to underdetermined systems with many more columns than rows. ParU does not solve such systems anyway but you might see some problems with other SuiteSparse packages.

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## 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 Linux/Unix

Before you compile the ParU library and demo programs, you may wish to edit the

SuiteSparse/SuiteSparse\_config/SuiteSparse\_config.mk configuration file. The defaults should be fine on most Linux/Unix systems and on the Mac. It automatically detects what system you have and sets compile parameters accordingly.

The configuration file defines where the BLAS libraries are to be found. Selecting the right BLAS is critical. There is no standard naming scheme for the name and location of these libraries. The defaults in the SuiteSparse\_config.mk file use -lblas; For best results, you should use the OpenBLAS at openblas.net (based on the Goto BLAS) [10], or high-performance vendor-supplied BLAS such as the Intel MKL, AMD ACML, or the Sun Performance Library. Selection of the BLAS is done with the BLAS= lines in the SuiteSparse\_config.mk file.

There are two parts that are important in chosing the compiler and the BLAS library.

'AUTOCC?= yes' This line let SuiteSparse\_config choose the compiler automatically. If there is an Intel compiler available it will be chosen. If you change yes to no then GCC will be used for the compilation.

'BLAS?= -lopenblas' This line let SuiteSparse\_config choose the BLAS library. By default ParU uses openBLAS. If you comment out this line ParU will look for the Intel Math Kernel Library.

After you decide about the compiler and BLAS library, type make at the Linux/Unix 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, and a set of simple demos will be run (including the one in the next section).

To test the lines of ParU, go to the Tcov directory and type make. To fully test the lines of ParU you should define PARU\_ALLOC\_TESTING and PARU\_COVERAGE in ParU\Source\paru\_internal.hpp. This will work for Linux only.

To install the shared library into /usr/local/lib and /usr/local/include, do make install. To uninstall, do make uninstall. For more options, see the SuiteSparse/README.txt file.

#### 2.2 C/C++ Example

The C++ interface is written using only real matrices. The simplest function computes the MATLAB equivalent of x=A\b and is almost as simple: 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 A and the residual. Some error testing code

is omited to simplify showing how the program works. The full program can be found in ParU/Demo/paru\_demo.cpp

```
#include "ParU.hpp"
int main(int argc, char **argv)
{
   cholmod_common Common, *cc;
   cholmod_sparse *A;
   ParU_Symbolic *Sym = NULL;
   //~~~~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~~~
   ParU_Control Control;
   ParU_Ret info;
   info = ParU_Analyze(A, &Sym, &Control);
   ParU_Numeric *Num;
   info = ParU_Factorize(A, Sym, &Num, &Control);
   double my_time = omp_get_wtime() - my_start_time;
         Test the results
   Int m = Sym -> m;
   if (info == PARU_SUCCESS)
   {
       double *b = (double *)malloc(m * sizeof(double));
       double *xx = (double *)malloc(m * sizeof(double));
       for (Int i = 0; i < m; ++i) b[i] = i + 1;
       info = ParU_Solve(Sym, Num, b, xx, &Control);
       printf("Solve time is %lf seconds.\n", my_solve_time);
       double resid, anorm;
       info = ParU_Residual(A, xx, b, m, resid, anorm, &Control);
       printf("Residual is |%.21f| and anorm is %.2e and roond is %.2e.\n",
              resid == 0 ? 0 : log10(resid), anorm, Num->rcond);
       free(b);
       free(xx);
   }
   Int max_threads = omp_get_max_threads();
   BLAS_set_num_threads(max_threads);
```

```
//~~~~~~Free Everything
ParU_Freenum(&Num, &Control);
ParU_Freesym(&Sym, &Control);

cholmod_l_free_sparse(&A, cc);
cholmod_l_finish(cc);
}
```

#### 2.3 C/C++ Syntax

ParU\_Ret is the output structure of all ParU routines. The user must check the output before continuing and computing further the result of prior routine. You can see the user callable routines in ParU/Include/ParU.hpp. The following is a list of user-callable C++ functions and what they can do:

- 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 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 stair case structure) matrix, computing factors and permutations is 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\_Freenum: frees the numerical part of factorization.
- 6. ParU\_Freesym: frees the symbolic part of factorization.

# 2.4 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 following files:

- 1. SuiteSparse/ParU/Include/ParU.hpp describes each C++ function. Only double and square matrices are supported.
- 2. SuiteSparse/ParU/Include/ParU.h describes the C-callable functions.

There are C/C++ options to control ParU which is an input argument to several routines. When you make ParU\_Control object it is initialized with default values. The user can change the values. Here is the list of control options:

ParU_Control	default value	explanation
mem_chunk	1024 * 1024	chunk size for memset and memcpy
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 max_row
panel_width	32	width of panel for dense factorization
paru_strategy	PARU_STRATEGY_AUTO	default strategy for ParU
piv_toler	0.1	tolerance for accepting sparse pivots
diag_toler	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
paru_max_threads	0	initialized with omp_max_threads

The first row of the options is used in the symbolic analysis. In the symbolic analysis phase, only the pattern of the matrix are probed. The second row of control options show those that has an impact on numerical analysis.

paru\_max\_threads is initalized 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 UMF-PACK, however the user can ask UMFPACK for a unsymmetric strategy but use a symmetric strategy for ParU. Usually UMFPACK choses a good ordering, however there might be cases that user 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 BLAS library must be chosen.

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 author for details.

#### References

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